

T-2914

ITERATIVE STACK IN PETROLEUM SEISMOLOGY

by

CHIE-CHUNG YEN

ARTHUR LAKES LIBRARY
COLORADO SCHOOL OF MINES
GOLDEN, COLORADO 80401

ProQuest Number: 10782594

All rights reserved

INFORMATION TO ALL USERS

The quality of this reproduction is dependent upon the quality of the copy submitted.

In the unlikely event that the author did not send a complete manuscript and there are missing pages, these will be noted. Also, if material had to be removed, a note will indicate the deletion.



ProQuest 10782594

Published by ProQuest LLC (2018). Copyright of the Dissertation is held by the Author.

All rights reserved.

This work is protected against unauthorized copying under Title 17, United States Code
Microform Edition © ProQuest LLC.

ProQuest LLC.
789 East Eisenhower Parkway
P.O. Box 1346
Ann Arbor, MI 48106 – 1346

T-2914

A thesis submitted to the Faculty and Board of Trustees of the Colorado School of Mines in partial fulfillment of the requirements for the degree of Master of Science (Geophysics).

Golden, Colorado

Date Apr. 19, '84

Signed: Yun Chiechung
Chie-chung Yen

Approved: Frank Hadsell
Frank Hadsell,
Thesis Advisor

Golden, Colorado

Date 27 April 84

Phillip R. Romig
Phillip R. Romig,
Head of Department
of Geophysics.

ABSTRACT

Recently, an iterative stack algorithm has been introduced by Naess (1979, 1981, 1982). The objective of this paper is to appraise its performance. It was checked in CDP-stack and in velocity analysis; both proved this is a powerful noise suppression tool.

After realizing its effectiveness, investigations were made to see its compatibility to other noise suppression tools. It seems non-compatible with the diversity stack or f-k filtering; which, should be avoided when the iterative stack is used.

Wiener deconvolution could not achieve its goal when executed after the iterative stack. Which infers, when possible, linear processes should precede non-linear processes.

Muting and manual editing will still benefit the iterative stack. In real data, a proper number of iterations should be carefully chosen to get the best enhanced signal.

TABLE OF CONTENTS

ABSTRACT.....	iii
TABLE OF CONTENTS.....	iv
LIST OF ILLUSTRATIONS.....	vi
ACKNOWLEDGEMENTS.....	ix
1. INTRODUCTION.....	1
2. ALGORITHMS REVIEW.....	3
2.1 Straight Stack.....	3
2.2 Weighted Stack.....	4
2.3 Stack after Coherence Picking.....	8
2.4 Stack after Maximum Likelihood Filtering or Optimum Deconvolution Filtering.....	9
2.5 Median Stack.....	12
2.6 Iterative Stack.....	12
3. SYNTHETIC EXAMPLES OF THE USE OF THE ITERATIVE ALGORITHM.....	19
3.1 Naess' Model - Example of iterative Stack.....	19
3.2 Naess' Model with Background Noise Involved.....	25
3.3 Example of Single Trace Process.....	29
3.4 Example of Velocity Analysis.....	33
3.5 A Discussion with More Realistic Modelling Work.	40
4. COMPARISON WITH THE OTHER ALGORITHMS.....	44
4.1 Comparison between the Straight, Weighted, and Iterative Stack in Suppressing Random Noise.....	44
4.2 Comparison between the Iterative Stack and the	

Velocity Filter in Suppressing Surface Travelling Noise and Multiples.....	57
4.3 Comparison between the Iterative Stack and Wiener Deconvolution in Suppressing Periodic Noise.....	69
5. CONCLUSIONS.....	87
6. REFERENCES.....	92
7. APPENDICES.....	94
APPENDIX A: MAIN PROGRAMS.....	94
APPENDIX B: SUBROUTINES.....	108
APPENDIX C: Output of Program 'GRAY'.....	141

LIST OF ILLUSTRATIONS

Figure	Page
2-1 Iterative algorithm illustrated by 10 random numbers.....	16
2-2 Iterative algorithm illustrated by 10 positive random numbers.....	17
3-1 A 3-layer model and its parameters.....	20
3-2 CDP-gather generated from the 3-layer model....	21
3-3 CDO-gather of figure 3-2 after NMO correction..	23
3-4 Result of iterative stack from figure 3-3.....	24
3-5 CDP-gather generated from 3-layer model with background noise involved.....	26
3-6 CDP-gather of figure 3-5 after NMO correction..	27
3-7 Result of iterative stack from figure 3-6.....	28
3-8 Result of single trace process.....	32
3-9 A 4-layer model and its parameters.....	34
3-10 CDP-gather generated from the 4-layer model....	35
3-11 Conventional constant velocity stack (CVS).....	37
3-12 Iterative CVS-analysis after 2 iterations.....	38
3-13 Iterative CVS-analysis after 4 iterations.....	39
3-14 12-fold CDP-gather generated from the 3-layer model by GRAY suite.....	42
4-1 Stack from different stacking algorithms with very high S/N ratio.....	46
4-2 Stack from different stacking algorithms with high S/N ratio.....	47

Figure	Page
4-3 Stack from different stacking algorithms with moderate S/N ratio.....	48
4-4 Stack from different stacking algorithms with low S/N ratio.....	49
4-5 Stack from different stacking algorithms with poor S/N ratio.....	50
4-6 Stack from different stacking algorithms with moderate S/N ratio but varied signal scale.....	51
4-7 Stack from different stacking algorithms with moderate S/N ratio but varied signal scale.....	52
4-8 Stack from different stacking algorithms with moderate S/N ratio but varied signal scale.....	53
4-9 Stepout/bandpass relation in f-k domain for CDP-gather traces with sampling rate of 2 ms...	59
4-10 Modelling traces generated from figure 3-1 with horizontally travelling noise added, already NMO corrected.....	60
4-11 Reject zone in f-k domain, upper: Mask for figure 4-12; lower: Mask for figure 4-13.....	62
4-12 Comparison of the f-k filtering result by a-half plane zero-out (left) and the 5-iteration of the iterative stack (right).....	63
4-13 Comparison of the f-k filtering result by 45 degree zero-out (left) and the 5-iteration of the iterative stack (right).....	64
4-14 Left shows the result of iterative process applied to the output of f-k filtering (half plane zero-out), right picture is the result of direct iterative stack.....	67

Figure	Page
4-15 Left shows the result of iterative process applied to the output of f-k filtering (45 degree zero-out), right picture is the result of direct iterative stack.....	68
4-16 12-fold CDP gather generated from the model in figure 3-1 with ghost added.....	72
4-17 Stacks by iterative process on the modelling CDP-gather of figure 4-16.....	74
4-18 The result of Wiener process on figure 4-17....	75
4-19 The result of Wiener process on figure 4-17 with Wiener operator designed against the second reflection.....	77
4-20 Iteration after muting.....	79
4-21 The result of Wiener process on figure 4-20....	80
4-22 The final iterative stacks made by single-trace-processing plus a muting function.....	81
4-23 The result of Wiener process on figure 4-22....	82
4-24 The result of predictive decon on figure 4-22..	83
4-25 Prestack deghosting result.....	85
4-26 The final section after iterative process on figure 4-25.....	86
 Table	
3-1 The rays generated by GRAY suite from the receiver #9 of the 3-layer model.....	43

ACKNOWLEDGMENTS

I would like to express my sincere thanks to my thesis advisor Dr. F. A. Hadsell who has provided helpful suggestions throughout this work, and to Mr. F. H. Ngian and Dr. W. A. Schneider who guided me through graduate studies as my committee members.

For financial support in form of a grant during the year of 1979, I am thankful to the National Science Council, Republic of China. For support in form of a graduate research assistantship from 1983 to present through the Geophysics Department of Colorado School of Mines, I am grateful.

My thanks also goes to my wife Kwaydong, who has given me moral support when I needed it. Last, but not least, I would like to thank my fellow Geophysics graduate students who offered constructive criticism and encouragement.

1. INTRODUCTION

A stack is a composite record made by mixing traces from different records. Examples in seismic exploration are ground mixing, instrument mixing, vertical stacking, uphole stacking, common depth point (CDP) stacking, common offset stacking, coherency filtering, diversity stacking, migrating by summation method, etc. (Sheriff 1982). The purpose of "stacking" is to enhance the signal-to-noise ratio in the final stacked trace.

The stacking algorithms commonly used in the above processes can be classified into five categories: straight summation, weighted summation, summation after coherency measure, summation after optimum filtering, and non-linear stacking methods. We have a brief review of these algorithms in the next chapter.

Recently, a non-linear iterative stacking algorithm has been introduced and described by Naess (1979). The algorithm has been shown to be adaptable to CDP stack and velocity analysis (Naess and Bruland 1981, and Naess 1982). The primary objective of this paper is to check the performance of the iterative stack in these applications, and to find way to put the iterative stack into routine seismic data processing procedures (All computer

programming work of this paper is assembled in
Appendices).

2. ALGORITHMS REVIEW

A quick review of stacking algorithms will be shown below in this order: straight stack, weighted stack, stack after coherency picking, stack after maximum likelihood processing, stack after optimum deconvolution filtering, median selection, and iterative stacking.

2.1 Straight Stack

The same weight is attached to each sample entering the straight stack. The process of straight stack of CDP NMO-corrected data has been well known since Mayne (1962). The enhancement of the S/N ratio by straight summation was an intuitive end result to him at that time. Robinson (1970) gave a quantitative argument that \sqrt{n} is the improvement of the S/N rms amplitude ratio for the straight summation from n traces if all traces have identical S/N ratios and identical signal amplitudes.

Straight stack is still the simplest version accepted as standard routine processing for seismic data. The disadvantage is that small but consistent signals may easily be overruled by a few large noise pulses.

2.2 Weighted Stack

S/N ratio will decrease with increased offset beyond some minimum offset. George et al (1968) suggested that a calculated factor based on offset distance should be multiplied with each trace before stacking.

Statistically optimum stacking was first proposed by Robinson (1970), a theory for weighting seismic records in the stacking process was developed from a statistical model, in accord with that model, a seismic record is described by its signal scale and its S/N energy ratio.

The following statistical properties are imposed on the signal and noise components within a simple seismic window by the model:

- 1) the signal is identical (except the scale) and correlated on all traces,
- 2) the noise on any given trace has zero mean and is not correlated to the noise on any other trace; and
- 3) the S/N ratio is reasonably constant for each individual trace over the time extent of the window.

The mathematical expression for the j th sample of the i th seismic trace in a gather consisting of I CDP traces is

$$t_{ij} = a_i (s_j + n_{ij}), \quad i=1,2,\dots,I$$

$$j=1,2,\dots,J \quad (1)$$

Where a_i is the signal amplitude factor or scale (arbitrarily, $a_i = 1$), s_j is the signal component, n_{ij} is the noise component, and J is the number of discrete samples per trace. The signal energy and the noise and total energies respectively for the i th trace are defined as follows:

$$S = \sum_{j=1}^J (s_j)^2, \quad N_i = \sum_{j=1}^J (n_{ij})^2, \quad T_i = \sum_{j=1}^J (t_{ij})^2 \quad (2)$$

The unique S/N energy ratio r for the i th trace is defined as

$$r_i = \frac{S}{N_i} \quad (3)$$

The objective of the optimum stack is to determine a weighting factor w_i to be applied to the corresponding i th trace such that the S/N ratio of the trace resulting from the weighted stack will be maximized.

The optimum stack for I traces is achieved by a logical extension of the two-fold weight stack: (see Robinson 1970, equations 4 to 17)

$$(t_j^{(I)})_{\text{optimum}} = \frac{1}{c} \sum_{i=1}^I \frac{r_i}{a_i} t_{ij} \quad (4)$$

where c is a normalization factor. Stated in words, the

optimum stack is effected by first equalizing the signal scale on each trace, then weighting the resulting traces with their respective signal-to-noise energy ratios, and finally, stacking and normalizing these traces.

To implement the optimum stack, we should know the scale and S/N ratio for each trace in the stack. We cannot determine these values in practice. The statistical estimations are based on the simple trace-cross-product sums defined as

$$x_{mn} = \sum_{j=1}^J t_{mj} t_{nj} , \quad (m,n=1,2,\dots,I) \quad (5)$$

Robinson (1970, equations 21 to 24) demonstrates that the best approximations to the trace scale and the S/N ratio for trace n in I traces are:

$$a_n = \frac{1}{(I-2)} \sum \frac{x_{nk}}{x_{1k}} , \quad \text{where } k \neq n$$

$$r_n = \frac{1}{(I-1)} \sum \frac{1}{\frac{x_{mm}x_{nn}}{x_{mn}} - 1} \quad (6)$$

The computation of these is very time-consuming. If we assume all S/N ratios are approximately equal (Robinson 1970, scheme IV), then approximately,

$$w_n \propto \left(\frac{1}{x_{nn}} \right)^{0.5} , \quad n=1,2,\dots,I . \quad (7)$$

This is a much simpler weighting function which merely effects a normalization of each seismic trace on its total energy. These weights are used in what is called the diversity stack.

In routine processing, bad traces are first deleted by manual editing . This gives bad traces a zero weighted value and assures that the S/N ratio of each trace falls in an acceptable range. The traces are then stacked by diversity stack. This simple combined process has been phenomenally consistent at achieving a notable improvement in the stack (Ngian, 1983).

2.3 Stack after Coherence Picking

Coherency stack combines data which satisfy certain trace-to-trace coherence criteria. Many coherence measures were reviewed by Neidell and Taner(1971). The common objectives of coherency measure include signal prediction, detection of shifts, and extraction of common signal. Signal in the channels can be measured, enhanced, shaped, or even ignored after the use of some coherence measure. It is too broad a subject for discussion here.

The coherency measure used in stack is largely dependent on the coherency criteria used. Coherence picking may be better than the manual editing in some sense, and the result of the coherent stack might be slightly better than that of a routine weighted process.

2.4 Stack after Maximum Likelihood Filtering or Optimum Deconvolution Filtering

Optimum stacking filters and their performance in suppressing uncorrelated noise has been investigated by White(1977). He defined the "optimum stack filtering" in terms of two processes; the "maximum likelihood process" which minimizes the output noise power under the restriction of no signal distortion, the "Wiener deconvolution process" which minimizes the mean square deviation of the output from some arbitrarily selected "desired" output.

The basic principles of maximum likelihood processing are still contained in the theory for the weighted stack, and the maximum likelihood weights w_i for the trace i or I CDP traces are obtained by:

$$w_i = \frac{1}{a_i} \cdot \frac{r_i}{\sum_{j=1}^I r_j} \quad (8)$$

where a_i is the signal scale and r_i is the S/N power ratio (see White 1977, equations 1-7).

Allowing for this leads to the case of optimum stacking filters, the above equation must be valid at every frequency in the signal

$$w_i(f) = \frac{r_i(f)}{a_i \cdot \sum_{j=1} r_j(f)} \quad (9)$$

Thus if, say, 17 frequencies between zero and the Nyquist frequency are allowed for the definition of the stack filters, the S/N ratios at these 17 frequencies must be evaluated to obtain the weights $w_i(f)$ from the above equation. The frequency response $w_i(f)$ can then be Fourier transformed to produce a 33 point stacking filter for trace i .

White (1977) concluded that optimally filtered stack is only slightly better than the straight stack. The improvement or S/N ratio of the straight stack will usually be at least eighty per cent of that of an optimally filtered stack. Moreover, the straight stack suppression of uncorrelated noise is best at low S/N ratios where it is most needed.

The multichannel Wiener filtering, on the other hand, actually requires only single channel operations. According to Green, Kelly and Levin(1966), Wiener multichannel filtering is essentially equivalent to the application of a single channel Wiener filter to the output of a maximum likelihood process. The later is more computationally effective than the former.

In fact, no matter whether one applies the maximum

likelihood filtering or the multichannel Wiener filtering, the improvement of the output will be restricted by the design parameters which must be obtained from the seismic data itself, and the best result is at most the desired trace (Ngian, 1983). This might be the reason the optimum stacking filters have not been widely used in the routine seismic data processing.

2.5 Median Stack

The median stack algorithm is to select only one "median-value" instead of combining all data. In the selection process, all data contributing to the stack are line-up in an order from high to low, and the one located in the middle will be chosen as the representative value (Ngian 1983).

This method is not linear and tends to change the frequency content in an unwanted manner (Naess 1979). It might be able to suppress the noise when a few large noise bursts overrule the small consistent signals. However, the iterative algorithm introduced in the next section will do a better job for the same price.

2.6 Iterative Stack

This algorithm has been first introduced and described by Naess(1979). The algorithm is expressed mathematically below by an example of CDP stack. The input to the algorithm consists of all amplitudes at one reflection time-level of the NMO-corrected CDP-gather. Parameters are defined as follows:

a_i^k = positive amplitude number i after k -th iteration,

b_j^k = negative amplitude number j after k -th iteration,

m = number of positive amplitudes,

n = number of negative amplitudes,

$M = m + n$ (note that Naess(1979) used M as a parameter),

S_+ = sum of positive group divided by M ,

S_- = sum of negative group divided by M .

The 1st-iteration:

$$S_+^1 = \frac{1}{M} \cdot \sum_{i=1}^m a_i^1$$

$$S_-^1 = \frac{1}{M} \cdot \sum_{j=1}^n b_j^1$$

$$S^1 = S_+^1 + S_-^1 = \frac{1}{M} \left(\sum_{i=1}^m a_i^1 + \sum_{j=1}^n b_j^1 \right) \text{ --1-iteration output,}$$

the 2nd-iteration:

$$\text{if } a_i^1 > S_+^1 \quad \text{set } a_i^{(2)} = S_+^1$$

$$a_i^1 \leq S_+^1 \quad a_i^{(2)} = a_i^1$$

$$b_j^1 < S_-^1 \quad b_j^{(2)} = S_-^1$$

$$b_j^1 \geq S_-^1 \quad b_j^{(2)} = b_j^1$$

$$\text{then } S_+^{(2)} = \frac{1}{M} \cdot \sum_{i=1}^m a_i^{(2)}$$

$$S_-^{(2)} = \frac{1}{M} \cdot \sum_{j=1}^n b_j^{(2)}$$

$$S^{(2)} = S_+^{(2)} + S_-^{(2)} \text{ --2-iteration output,}$$

The higher-iterations just repeat the computation above, the

q -th iterative stack is

$$S_+^q = \frac{1}{M} \cdot \sum_{i=1}^m a_i^q$$

$$S_-^q = \frac{1}{M} \cdot \sum_{j=1}^n b_j^q$$

$$S^q = S_+^q + S_-^q \text{ --}q\text{-iteration output.}$$

It is noted that the output of the 1st-iteration is equivalent to the result of straight stack.

This is a procedure where the difference between each input trace and the mean value is subtracted from the input trace; the separation of the positive and negative group is to make the subtracting procedure convergent. To put it another way, if the samples are not separated the process that causes the positive values to converge will cause the negative values to diverge.

The iterative algorithm could be considered as a combination of two algorithms:

- 1) an algorithm to find the sum of the most consistent positive amplitudes and the most consistent negative values,
- 2) an algorithm to enhance the most consistent polarity; for example, if all amplitudes are equivalent in absolute value, the computation formula could be represented by

$$S^q = (m/M)^q A + (n/M)^q B$$

where $a_1 = a_2 = \dots = a_m = A$

$$b_1 = b_2 = \dots = b_n = B$$

which could be named as "higher degree stacking".

Generally the primary amplitudes should be equal in strength and polarity, and all noise amplitudes shall be random or less consistent than primary ones. Thus, after several iterations the primary amplitudes will be emphasized

since the noise is suppressed more quickly.

In order to illustrate the iterative process, we made a test computation on a set of 10 random numbers as shown in figure 2-1. The initial 10 random numbers range between -0.5 and +0.5, six of them are positive. Since neither the positive group nor the negative group are dominant, both groups were suppressed quickly by iterative process. Negative group dropped to the zero level first (after 5 iterations) because they are relatively inconsistent with respect to the positive group. After 10 iterations the positive group was also exterminated.

Another test was made on a set of 10 positive random numbers which range between 0.00 and 1.00 (figure 2-2). The process was made up to 20 iterations. It clearly demonstrates the rate that data values were changing; when more data were falling near the same value (more consistency appeared) the slower their values would be decreased. This explains how the iterative stack sustains the small but consistent signals.

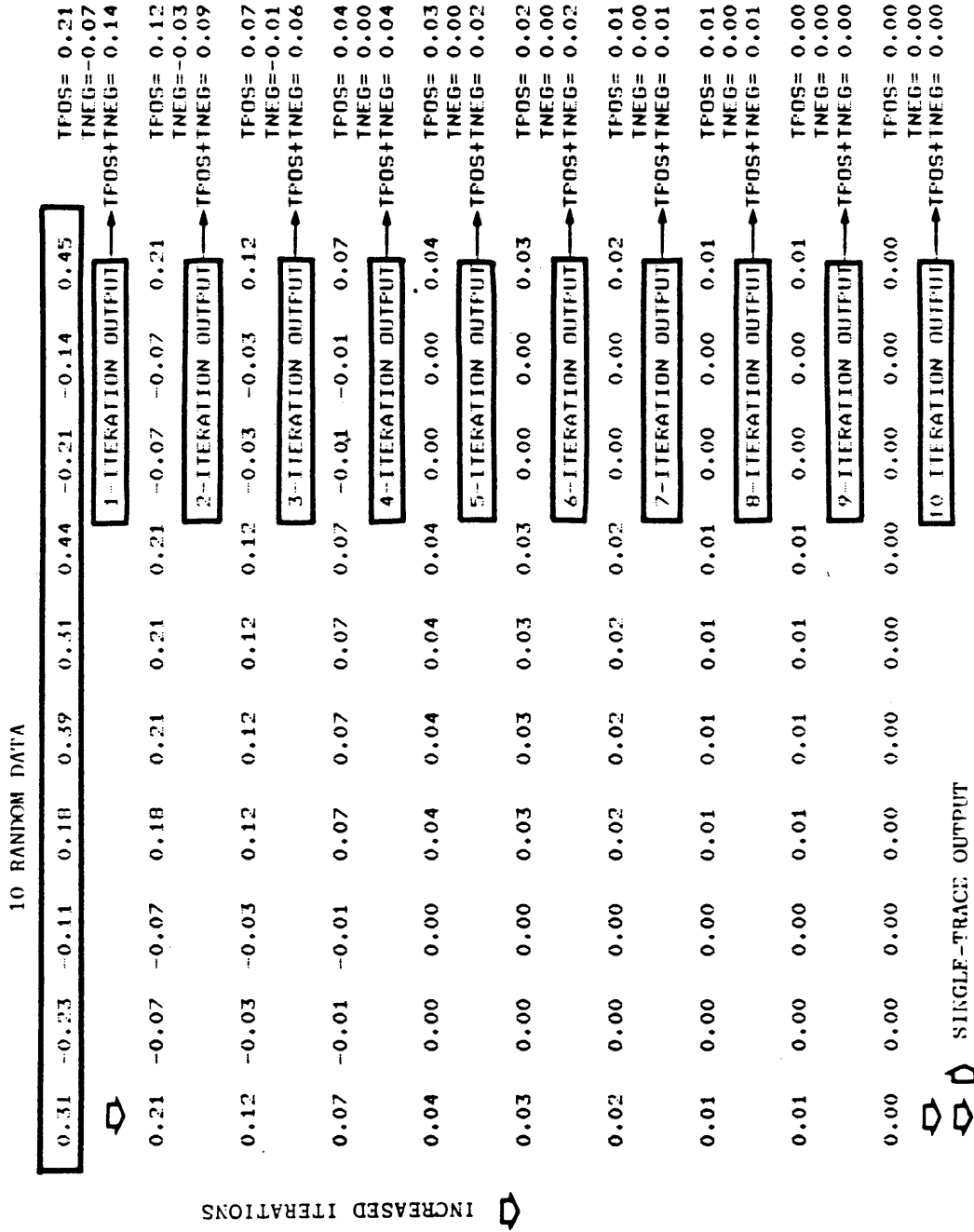


Figure 2-1. Iterative algorithm tested by 10 random numbers.

INCREASED ITERATIONS

0.19	0.73	0.61	0.32	0.11	0.19	0.06	0.71	0.64	0.05	TFOS= 0.36 TNEG= 0.00 ---TFOS+TNEG= 0.36
1-ITERATION OUTPUT										
0.19	0.36	0.36	0.32	0.11	0.19	0.06	0.36	0.36	0.05	TFOS= 0.24 TNEG= 0.00 ---TFOS+TNEG= 0.24
2-ITERATION OUTPUT										
0.19	0.24	0.24	0.24	0.11	0.19	0.06	0.24	0.24	0.05	TFOS= 0.18 TNEG= 0.00 ---TFOS+TNEG= 0.18
3-ITERATION OUTPUT										
0.18	0.18	0.18	0.18	0.11	0.18	0.06	0.18	0.18	0.05	TFOS= 0.15 TNEG= 0.00 ---TFOS+TNEG= 0.15
4-ITERATION OUTPUT										
0.15	0.15	0.15	0.15	0.11	0.15	0.06	0.15	0.15	0.05	TFOS= 0.12 TNEG= 0.00 ---TFOS+TNEG= 0.12
5-ITERATION OUTPUT										
0.12	0.12	0.12	0.12	0.11	0.12	0.06	0.12	0.12	0.05	TFOS= 0.11 TNEG= 0.00 ---TFOS+TNEG= 0.11
6-ITERATION OUTPUT										
0.11	0.11	0.11	0.11	0.11	0.11	0.06	0.11	0.11	0.05	TFOS= 0.10 TNEG= 0.00 ---TFOS+TNEG= 0.10
7-ITERATION OUTPUT										
0.10	0.10	0.10	0.10	0.10	0.10	0.06	0.10	0.10	0.05	TFOS= 0.09 TNEG= 0.00 ---TFOS+TNEG= 0.09
8-ITERATION OUTPUT										
0.09	0.09	0.09	0.09	0.09	0.09	0.06	0.09	0.09	0.05	TFOS= 0.08 TNEG= 0.00 ---TFOS+TNEG= 0.08
9-ITERATION OUTPUT										
0.08	0.08	0.08	0.08	0.08	0.08	0.06	0.08	0.08	0.05	TFOS= 0.08 TNEG= 0.00 ---TFOS+TNEG= 0.08
10-ITERATION OUTPUT										
0.08	0.08	0.08	0.08	0.08	0.08	0.06	0.08	0.08	0.05	TFOS= 0.07 TNEG= 0.00 ---TFOS+TNEG= 0.07
11-ITERATION OUTPUT										
0.07	0.07	0.07	0.07	0.07	0.07	0.06	0.07	0.07	0.05	TFOS= 0.07 TNEG= 0.00 ---TFOS+TNEG= 0.07
12-ITERATION OUTPUT										
0.07	0.07	0.07	0.07	0.07	0.07	0.06	0.07	0.07	0.05	TFOS= 0.07 TNEG= 0.00 ---TFOS+TNEG= 0.07
13-ITERATION OUTPUT										
0.07	0.07	0.07	0.07	0.07	0.07	0.06	0.07	0.07	0.05	TFOS= 0.06 TNEG= 0.00 ---TFOS+TNEG= 0.06
14-ITERATION OUTPUT										
0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.05	TFOS= 0.06 TNEG= 0.00 ---TFOS+TNEG= 0.06
15-ITERATION OUTPUT										
0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.05	TFOS= 0.06 TNEG= 0.00 ---TFOS+TNEG= 0.06
16-ITERATION OUTPUT										
0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.05	TFOS= 0.06 TNEG= 0.00 ---TFOS+TNEG= 0.06
17-ITERATION OUTPUT										
0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.05	TFOS= 0.06 TNEG= 0.00 ---TFOS+TNEG= 0.06
18-ITERATION OUTPUT										
0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.05	TFOS= 0.06 TNEG= 0.00 ---TFOS+TNEG= 0.06
19-ITERATION OUTPUT										
0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.06	0.05	TFOS= 0.06 TNEG= 0.00 ---TFOS+TNEG= 0.06
20-ITERATION OUTPUT										

Figure 2-2. Iterative algorithm tested by 10 positive random numbers.

Summary of the algorithm review:

(1) Straight summation is still the conventional routine stack algorithm, but actually it is an "optimum" weighted stack because manual editing and the diversity stack are usually added.

(2) Coherent picking might be better than the manual editing, but it is largely dependent on the coherency criteria used.

(3) Multichannel optimum filtering is unlikely to be as popular as straight stack, it is not adapted to routine multichannel processing.

(4) Iterative and median stack are non-linear processing, and hence change the frequency content.

3. SYNTHETIC EXAMPLES OF THE USE OF THE ITERATIVE ALGORITHM

We are going to use a simple model to check the performance of the iterative algorithm in the stacking of multichannel data, the algorithm will also be examined with background noise. The same model will be used to illustrate the performance of the iterative algorithm in single trace processing. A second model will be used to show the filtering effect of the iterative algorithm in velocity analysis. Finally we will introduce more realistic modelling for comparison and discussion.

3.1 Naess' Model- Example of Iterative Stack

A 3-layer earth model and its parameters is shown on figure 3-1. For this model I generated the 12-fold CDP-gather shown in figure 3-2. Each trace consists of a waterbottom reflection, two primaries, and five direct waterbottom multiples. The waterbottom reflection and all multiples were given the same strength. The primary-to-multiple amplitude ratio is assigned 1:10. The basic wavelet used in the model is shown in Appendix A: The program 'TRPLOT'. These model parameters are similar to

MODEL PARAMETERS

sample interval: 2 ms
primary-multiple amplitude ratio: 1:10
recording in 12-fold
minimum offset 100 m, maximum offset 650 m.
the model consists of waterbottom reflection, 2 primarys,
and 5 direct waterbottom multiples.

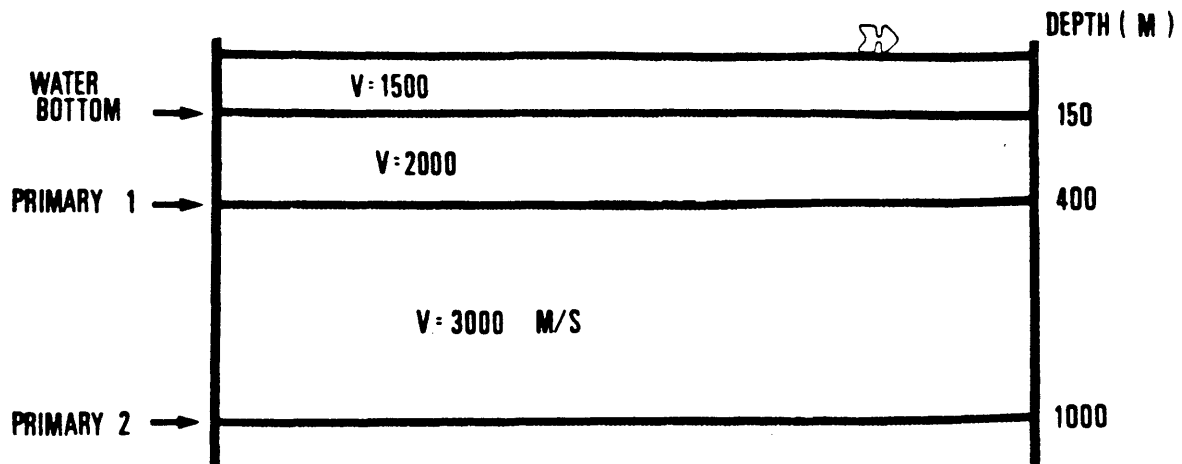


Figure 3-1, A 3-layer model and its parameters.

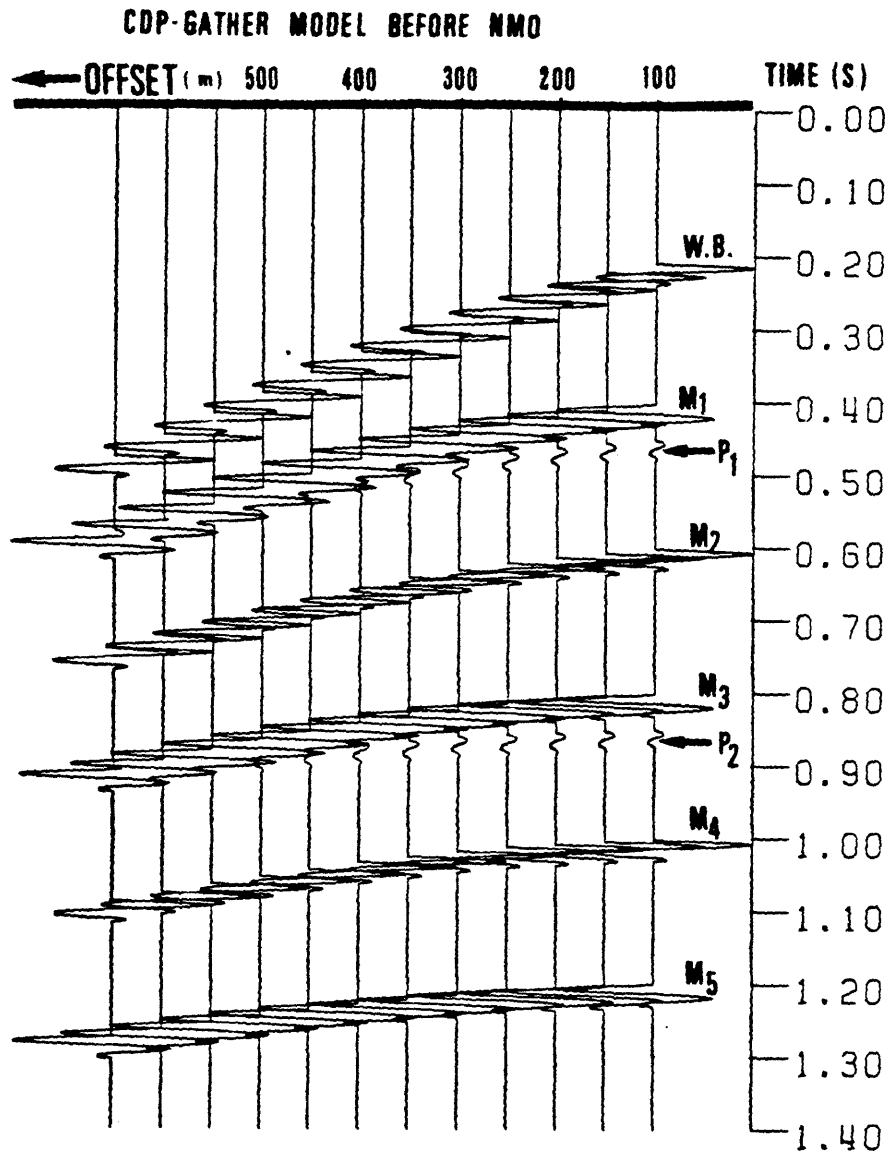


Figure 3-2, CDP-gather generated from the 3-layer model.

Naess' model (Naess 1979) and provide simplified reflections for easy observation. The realism of such data, as a model of a preprocessed gather, is discussed later in this chapter.

Figure 3-3 shows the CDP-gather traces after NMO-correction, the stretching effect is significant in far offset traces. Figure 3-4 displays the resulting final stacked traces after applying the iterative algorithm from 1-iteration to 12-iterations. The 1-iteration version is the same as straight stacking.

All the multiples are more and more reduced as the number of iterations increases. After five iterations the primary reflectors already dominate the trace. This test proves that the iterative stacking is a powerful tool and the 5-iteration or 6-iteration version is probably adequate for the display of the final seismic section.

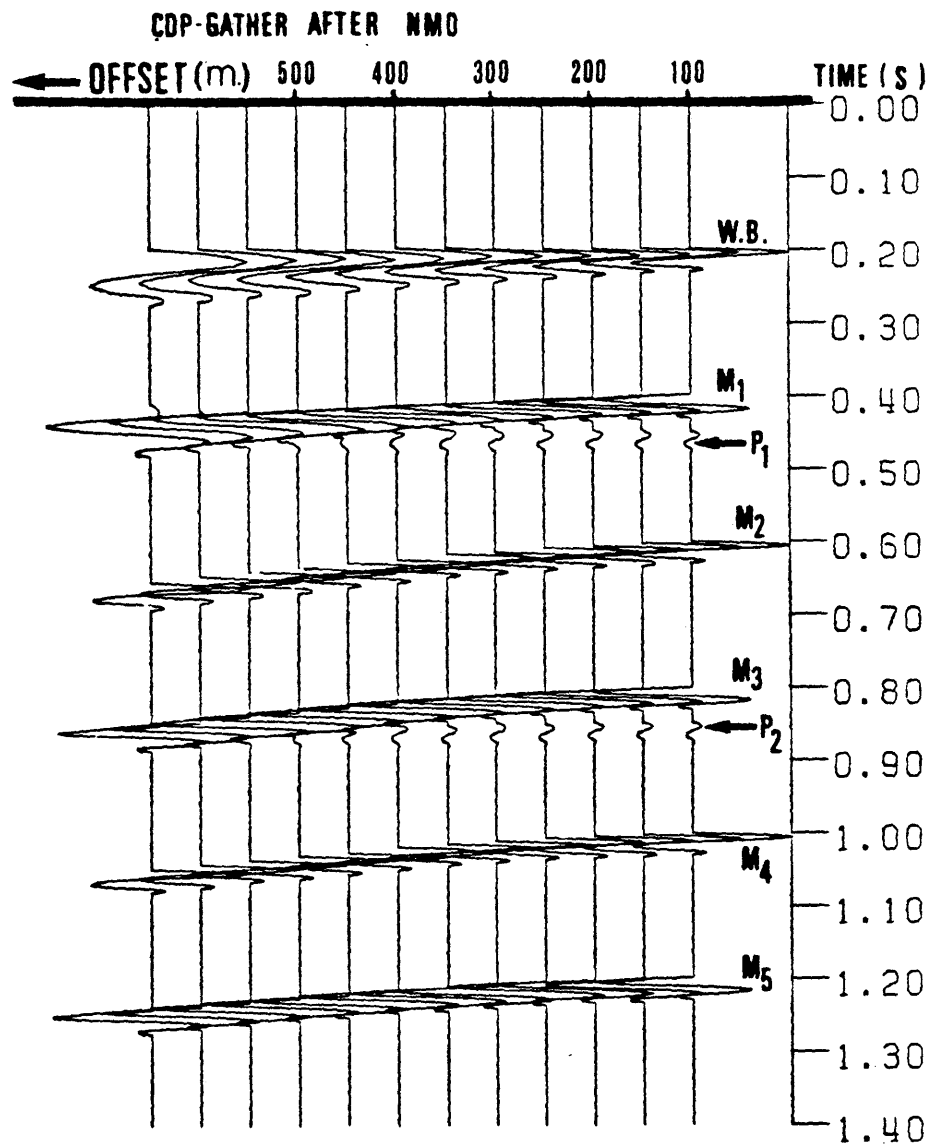


Figure 3-3, CDP-gather of figure 3-2 after NMO-corrected.

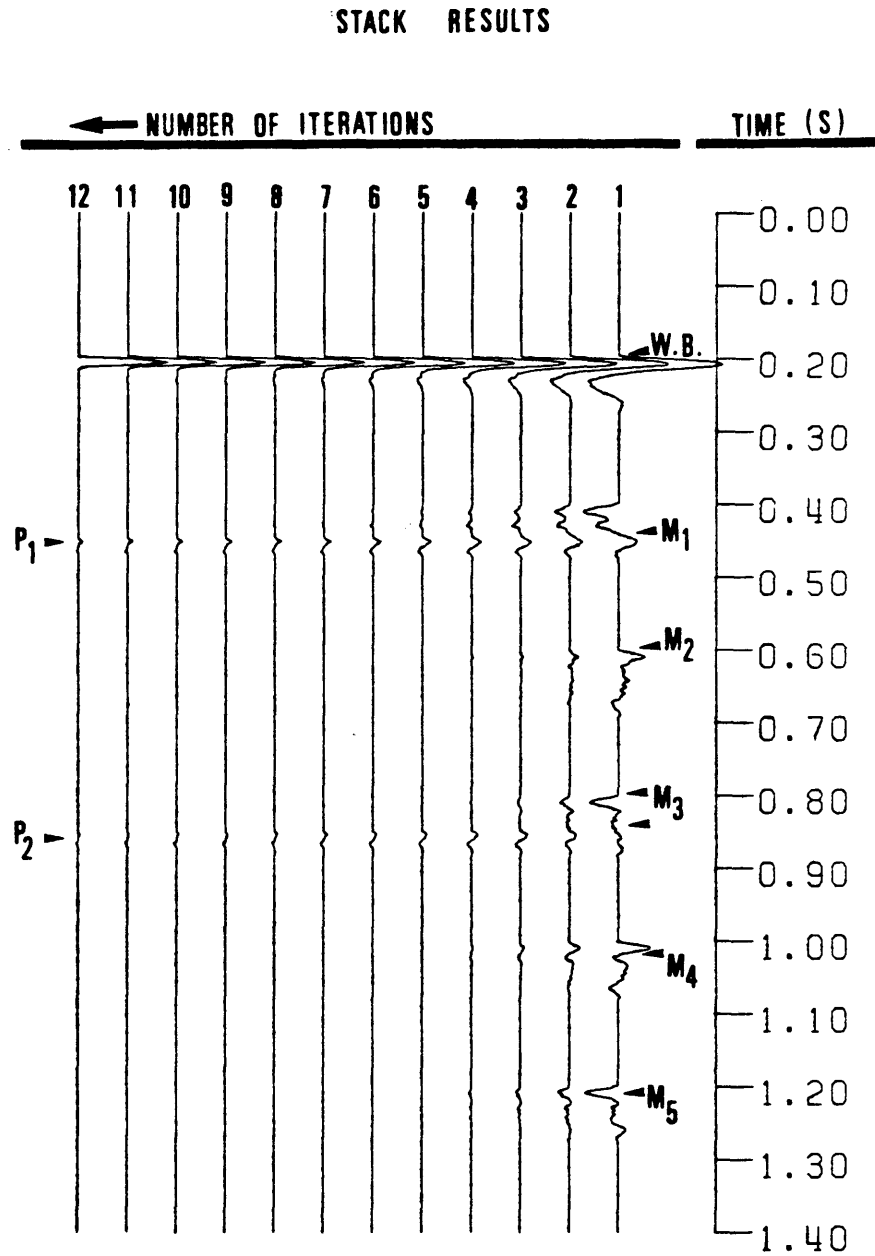


Figure 3-4, Result of iterative stack from figure 3-3.

3.2 Naess' Model with Background Noise Involved

It is interesting to know the performance of iterative stacking when background noise is involved in the model. The background noise is generated as filtered random numbers and added to each trace, the signal-to-noise amplitude ratio is about 5:1, where the signal value is calculated from the average of the three primaries, and the noise value is calculated from the average absolute value over the whole trace.

Figure 3-5 shows a 12-fold CDP-gather from the previous model with random noise added. Figure 3-6 shows NMO-corrected traces, and figure 3-7 displays the resulting final stack trace after 1-iteration to 12-iterations. Most of the background noise is removed from the stacked traces after 6 iterations. It could be stated that as long as the noise is not changing the consistency of the signal, as would be the case given a statics problem, the noise will be cleaned out of the section while the signal stays.

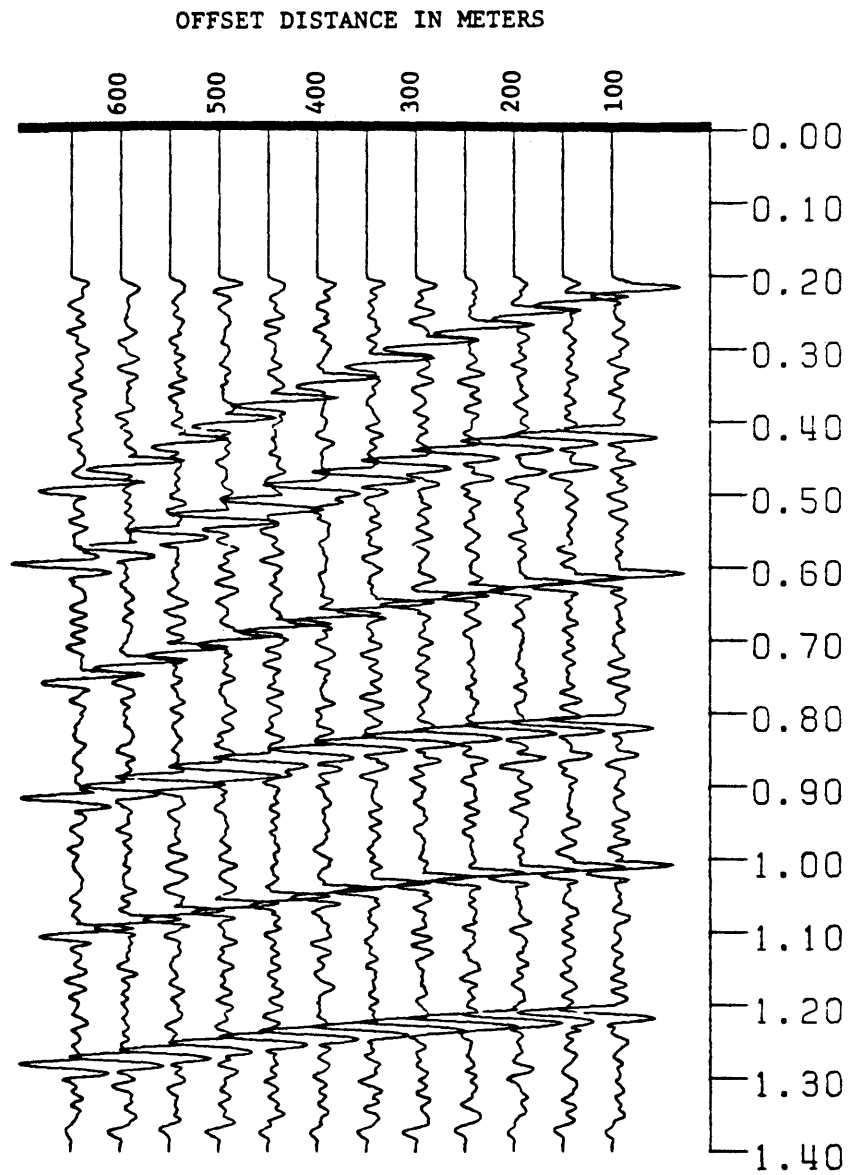


Figure 3-5, CDP-gather generated from 3-layer model with background noise involved.

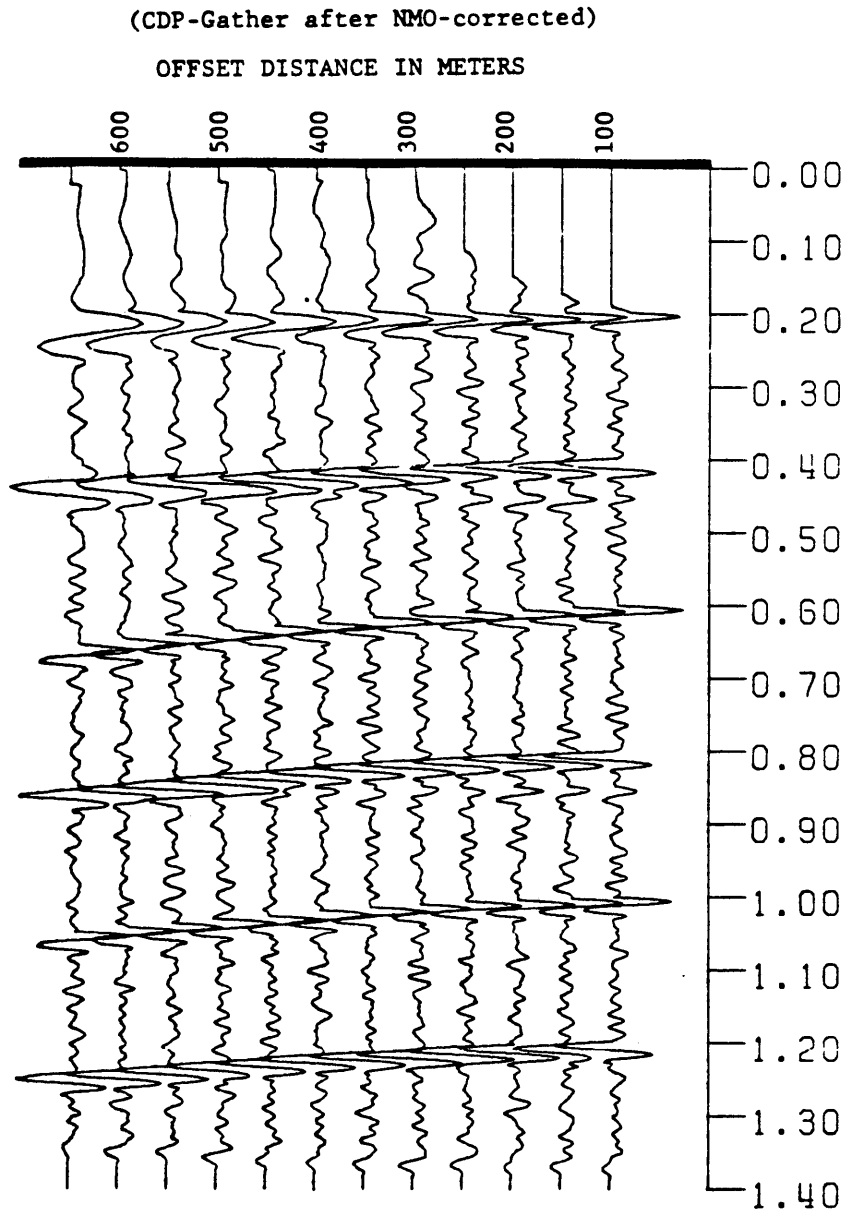


Figure 3-6, CDP-gather of figure 3-5 after NMO correction.

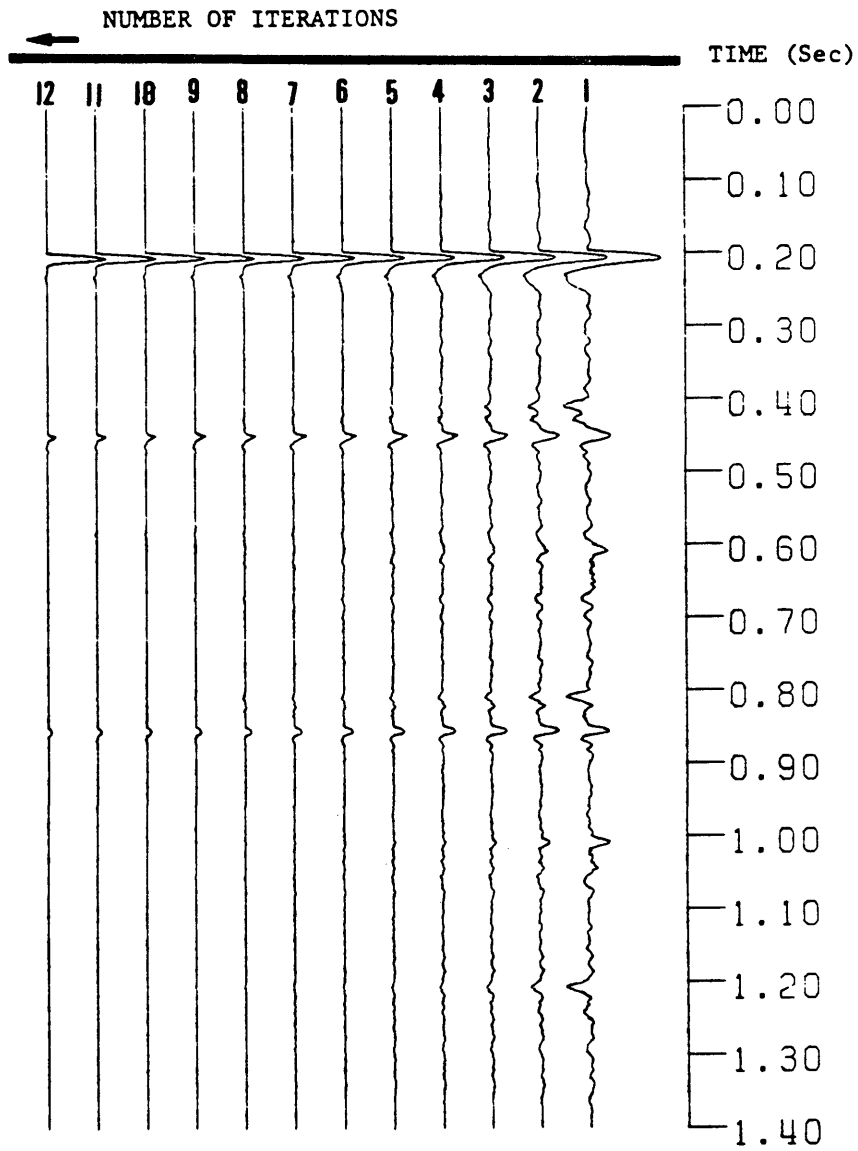


Figure 3-7, Result of iterative stack from figure 3-6.

3.3 Example of Single Trace Process

High resolution has always been a goal of the exploration industry. The maximum resolution in seismic reflection work would be obtained by using a point-source and recording the reflected signals with one single detector located as near to the source as possible. But the conventional CDP-stacking, whose sole object is to improve the signal-to-noise ratio and attenuate multiples, implies straight summation of signals recorded at distances up to 2.5 km from the source, there is a contradiction in using the CDP-method when high resolution is required.

The resolution in seismic reflection data may be subdivided into lateral and vertical resolution: the lateral resolution is concerned with the ability to detect small changes horizontally from trace to trace. In CDP-stacking, if the lateral extension of the reflecting area is larger than the extension of the change we want to detect, then the stacked data have a tendency to hide such a change. For example, a small fault may appear as a slight weakening in a reflector instead of a clear break.

The vertical resolution is concerned with minimum distance between two reflectors which is necessary in order to detect them as two separate events. The CDP-stacking

procedure affects the vertical resolution for two reasons: first, the stacking implies summation of traces with large offsets and hence more attenuation. Thus vertical resolution of a stacked trace is less than that of a single trace with short offset. Second, the stretching effect in normal moveout corrections may introduce further differences.

To obtain high resolution it would therefore be best to use only one trace recorded with the least offset for construction of the final seismic section. However, it is obviously also necessary to improve the S/N ratio on the near trace to a degree comparable to a conventionally stacked trace. A way towards achieving both of these goals is the use of iterative stacking with respect to the near trace only, that is called the single trace process (Naess, 1982). In practice, the iterative stacking is done equally with respect to all traces in the CDP-gather, and only the near trace is chosen as the output rather than the sum of the positive and negative groups.

Figure 3-8 displays the resulting final iterative stack with respect to the near trace from 1-iteration to 12-iterations, the 1-iteration shows more high frequency components than the conventional straight stack of figure 3-7, the difference of frequency content is also clearly evident when one compares the waterbottom reflections in the

traces on both figures. Furthermore, single trace processing shows a better ability to keep its original form than the ordinary iterative stack especially in the first few iterations.

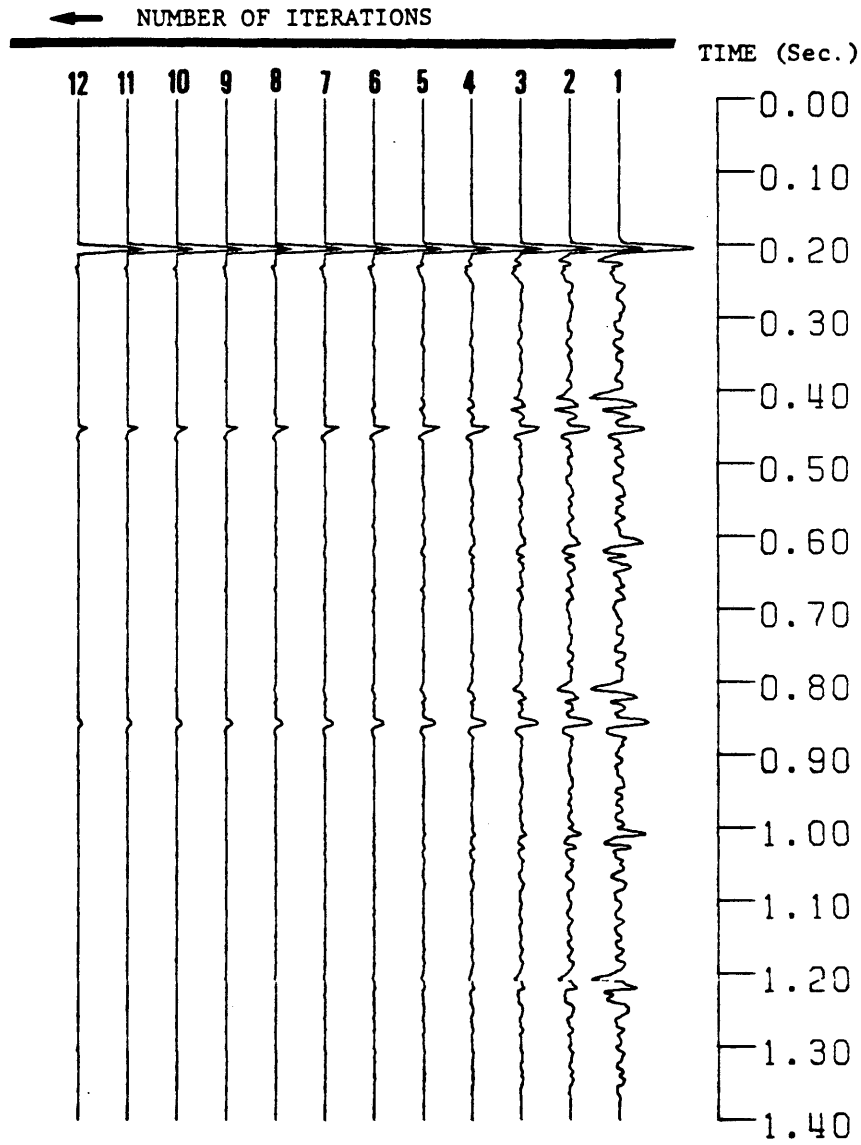


Figure 3-8, Result of single trace process.

ARTHUR LINDS LIBRARY
COLORADO SCHOOL of MINES
GOLDEN, COLORADO 80401

3.4 Example of Velocity Analysis

Velocity analysis is one of the important elements in seismic data processing since stacking velocities determine the quality of the final seismic sections. The velocities are also important independently of stacking, since they are related to physical parameters of geological significance.

Methods of velocity analysis are based on a visual or computer coherency measure. Prior to the coherency measurements there is normally no filtering except possibly a deconvolution, and the result of the velocity analysis is therefore highly dependent on the capability of the coherency measure to disregard the disturbing influence of noise events. Coherency measures are often based on different design philosophies (Neidell and Taner 1971), but the assumptions about noise and signal structures are seldom achieved in practice.

If disturbing noise can be removed from the data before the coherency measurements, the quality of velocity analysis will be improved. Naess (1981) has proposed that the iterative algorithm can achieve this filtering job, a check of its performance is illustrated below by a second model.

Figure 3-9 shows a 4-layer model and its parameters,

Synthetic CDP-gather Model

This is based on horizontally layered model with constant parameters within each layer and has 3 primary reflections P1,P2,P3 in addition to the waterbottom reflection, also water confined multiples are included.

recording in 12-fold, minimum offset 200 meters, maximum offset 1300 meters.

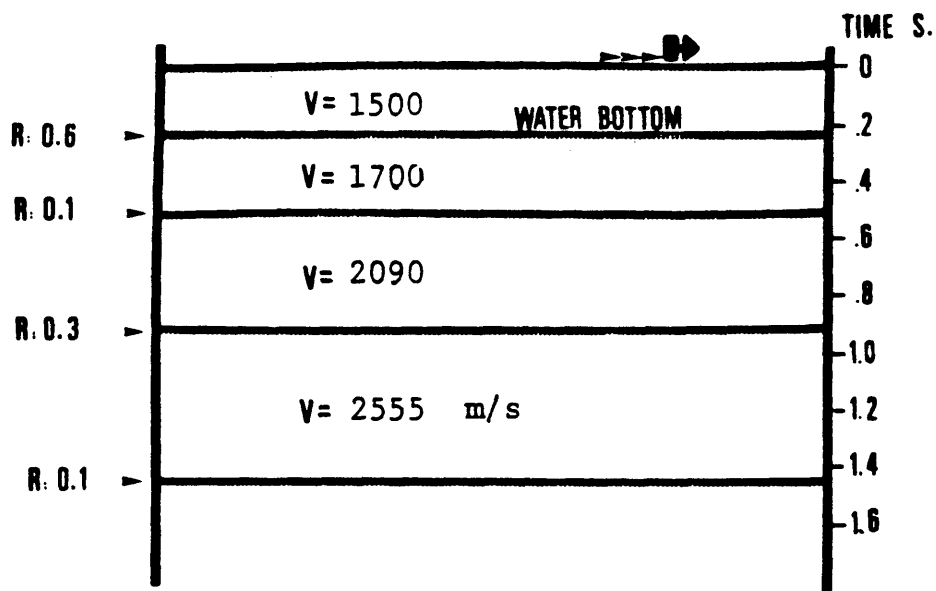


Figure 3-9, A 4-layer model and its parameters.

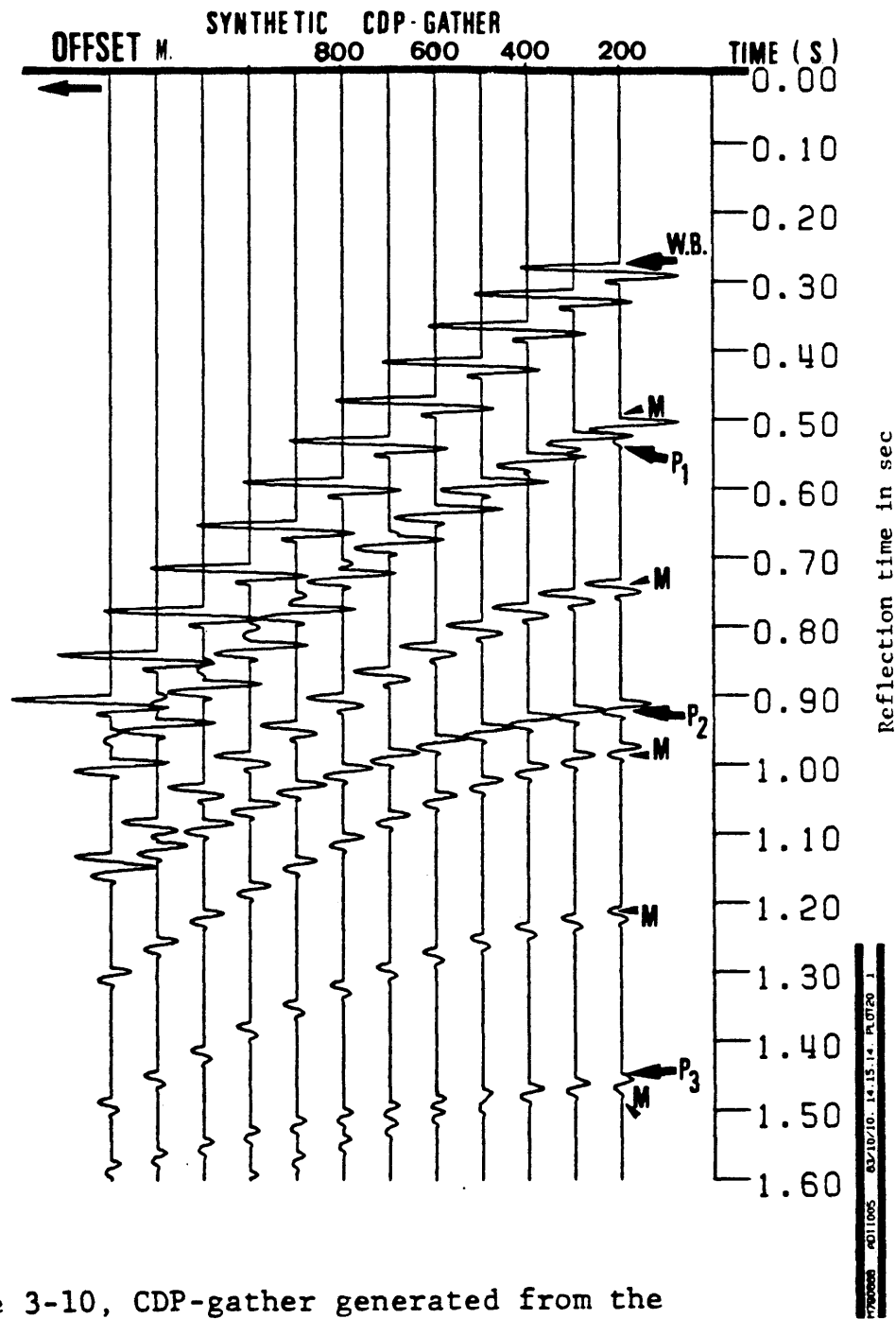


Figure 3-10, CDP-gather generated from the 4-layer model.

the CDP-gather shown on figure 3-10 is generated from this model under the assumption that only water confined multiples are important, these can be easily seen as a sequence of events following the primaries. The coverage is 12-fold, minimum offset is 200 m and maximum offset is 1300 m. The interval velocities used in the model are 1500, 1700, 2090, 2555 m/s respectively from top to bottom. The reflection coefficients used are 0.6, 0.1, 0.3, and 0.1 from water bottom and downwards. (See the next section for a discussion of the realism of the model.)

The method of velocity analysis we chose is the constant velocity stack (CVS) method. A conventional CVS-analysis is shown in figure 3-11, a display of iterative CVS-analysis after 2 iterations is shown on figure 3-12, and figure 3-13 shows an iterative CVS-analysis after 4 iterations. The velocities used range from 1460 m/s to 2240 m/s with an increment of 20 m/s. There is an impressive improvement in the definition of primary events on the iterative CVS versions. The background noise, caused by random line-ups which are present on the conventional CVS, have been effectively suppressed on the 4-iteration version.

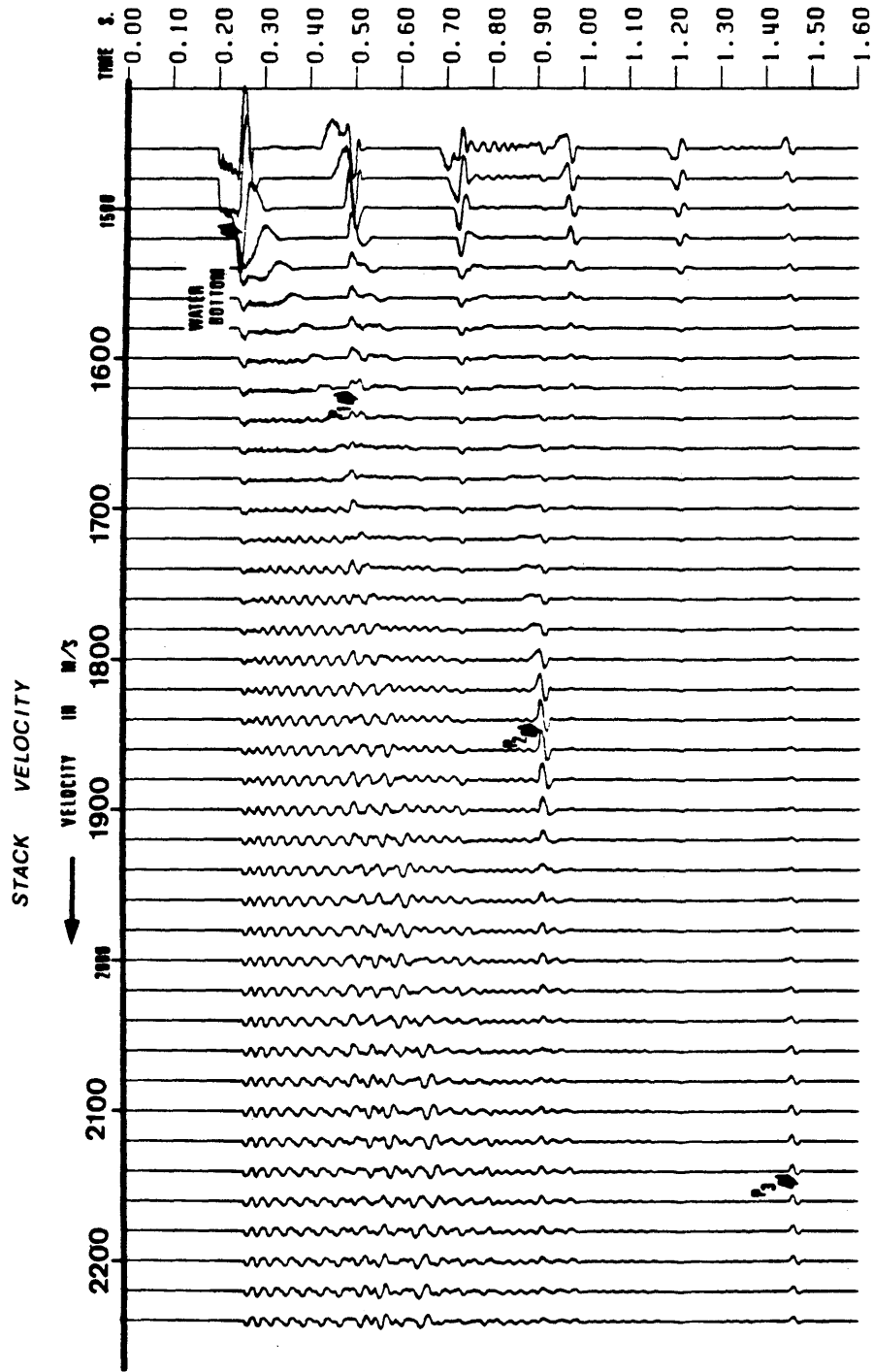


Figure 3-11, Conventional constant velocity stack(CVS).

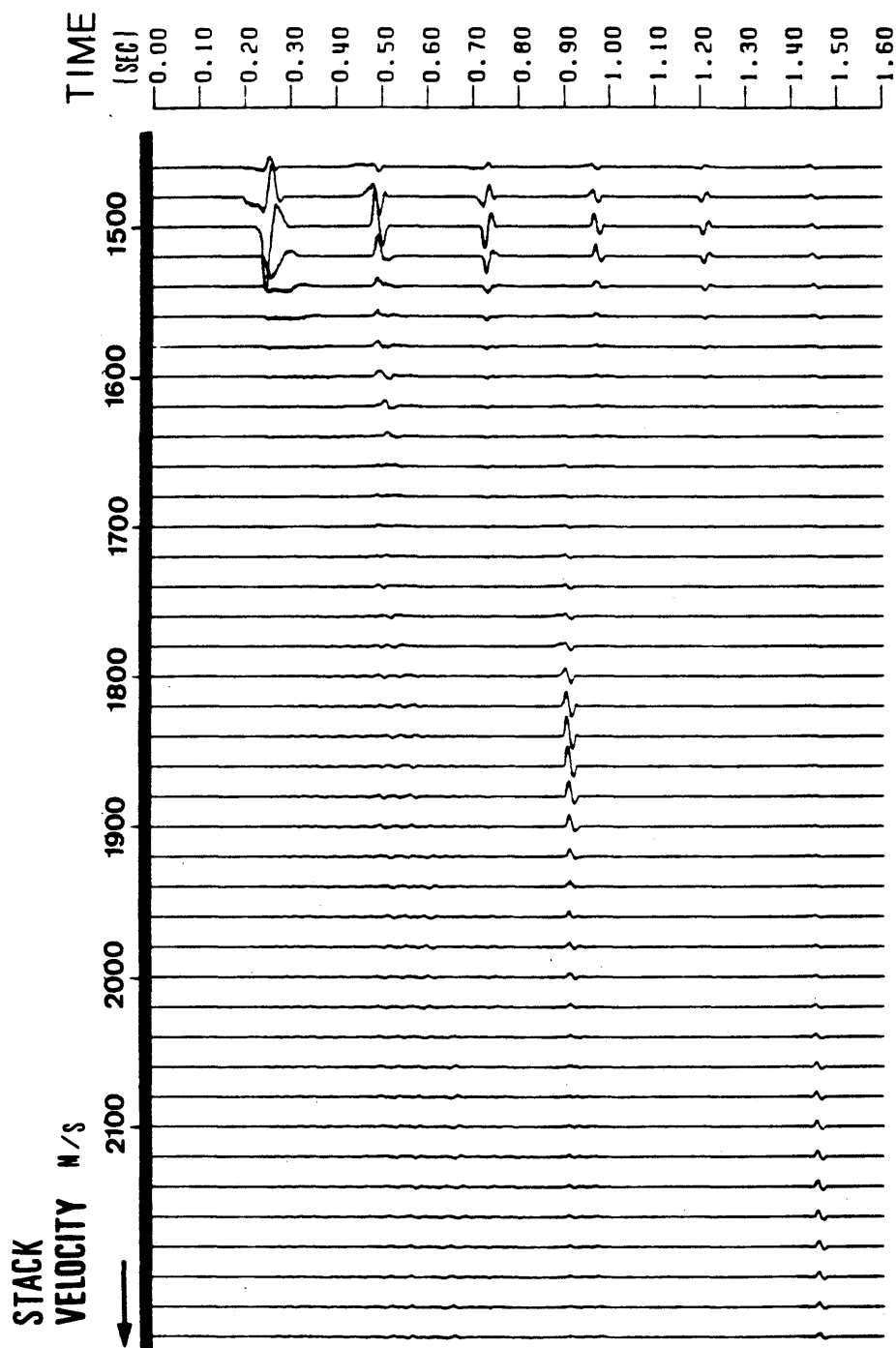


Figure 3-12, Iterative CVS-analysis after 2 iterations.

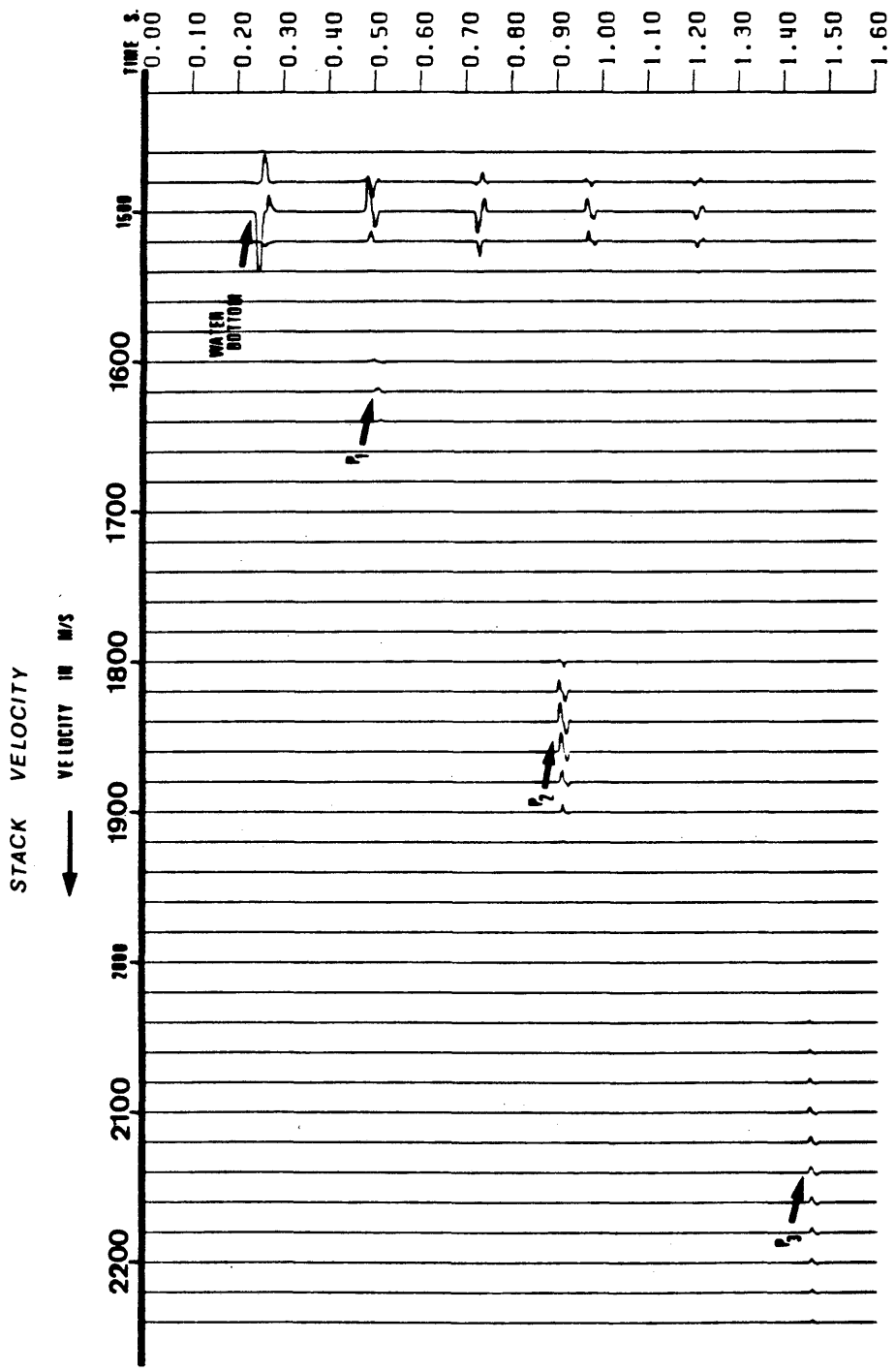


Figure 3-13, Iterative CVS-analysis after 4 iterations.

3.5 A Discussion with More Realistic Modelling Work

In the previous two modelling cases, the CDP-gather generated from the models were simplified in that only several easily observed raypaths were counted and reflection coefficients were not computed from the velocities. In order to justify these assumptions, a more realistic modelling suite of programs was applied. These are called the GRAY suite (Generalized RAY) of the Colorado School of Mines. The output of GRAY for the first model consists of 38 rays for each receiving station as shown in table 3-1, which is generated under the consideration that reflections come from both p-wave and s-wave ray-tracing with eight or less legs and 1.5 seconds recording time. The previous modelling work shown in figure 3-2 included only seven of these rays.

More rays means more disturbing events that hide the primaries. Figure 3-14 shows the more realistic CDP-gather from the GRAY suite, in which primaries are disturbed and are not so consistent as those in figure 3-2. The most interfering events are those strong responses from head waves and shear wave conversions in shallow depth as indicated in figure 3-14 and table 3-1.

The other important difference between our model and

the GRAY result is that the output of GRAY has taken into account the fact that the reflection amplitudes are changing with respect to the incidence angles. In the real earth, the changing does happen and the signal responses gathered from different offset angle could not be equal in strength or polarity as we have in our model, and hence the strength of the signal in the iterative stack will be also gradually decreased on iterations.

Therefore, we should carefully choose the number of iterations for the output of our iterative process. In our modelling test, for example, the number of iterations which gave the best display was 5 or 6; and it might be reduced to 3 or 4 in real data processing.

It is also suggested the muting function should be included in the process of the iterative stack. Naess (1979) commented that muting is not required because the stretching effect does not bother the iterative stacking too much. That is only based on his simple modelling result. If we consider the disturbing events from shallow paths and the fact that amplitude changes versus incidence angle, a proper muting will be highly helpful in regulating signal responses especially for shallow depths.

N LEGS:	9.	CLIP:	0.8279
SOURCE:	P		
S DEPTH:	0.01	G DEPTH:	0.01
R:	100.00	DZ:	0.00
		DR:	50.00
VP	VS	DEN	ZI
1.50	0.10	1.00	150.00
2.00	1.11	2.00	250.00
3.00	1.67	2.20	600.00
3.50	1.94	2.50	99.99

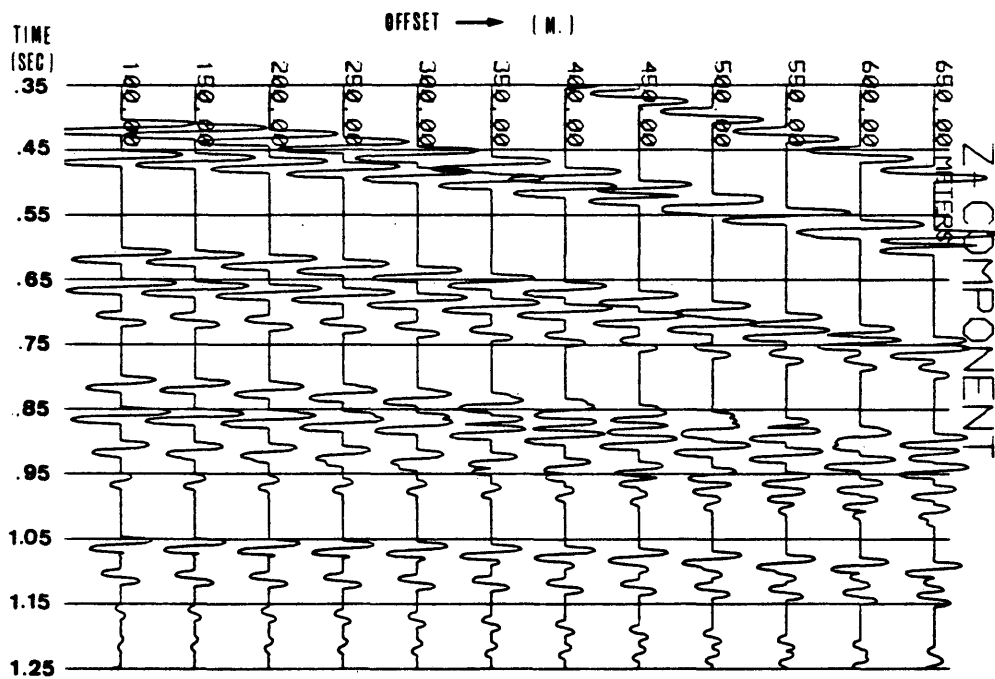


Figure 3-14, 12-fold CDP-gather from the 3-layer model generated by GRAY suite.

SPECIES	ARRIVAL TIME	VERTICAL COMPO.	RADIAL COMPO.	DIR	ND	RAYNAME
1	388723E+03	23527E+00	339235E+00	-	1	OPI
1	3887279E+03	287895E+00	479857E+00	-	1	HI
2	3887273E+03	1853926E+00	154454E+00	-	1	PI
3	388728E+03	1933827E+00	1344899E+00	-	1	PI
4	3887286E+03	847701E+00	44866E+00	-	1	PI
5	3887291E+03	853318E+00	2799687E+00	-	1	PI
6	3887293E+03	1053193E+00	3799879E+00	-	1	PI
7	3887291E+03	1244328E+00	4899239E+00	-	1	PI
8	3887294E+03	343333E+00	433879E+00	-	1	PI
9	3887298E+03	348424E+00	437813E+00	-	1	PI
10	3887298E+03	3444439E+00	468616E+00	-	1	PI
11	3887298E+03	3478033E+00	1783353E+00	-	1	PI
12	3887298E+03	3444433E+00	177613E+00	-	1	PI
13	3887298E+03	3444433E+00	36953E+00	-	1	PI
14	3887298E+03	3444433E+00	1788613E+00	-	1	PI
15	3887298E+03	3444433E+00	36782E+00	-	1	PI
16	3887298E+03	3444433E+00	33346E+00	-	1	PI
17	3887298E+03	3444433E+00	33346E+00	-	1	PI
18	3887298E+03	3444433E+00	33346E+00	-	1	PI
19	3887298E+03	3444433E+00	33346E+00	-	1	PI
20	3887298E+03	3444433E+00	33346E+00	-	1	PI
21	3887298E+03	3444433E+00	33346E+00	-	1	PI
22	3887298E+03	3444433E+00	33346E+00	-	1	PI
23	3887298E+03	3444433E+00	33346E+00	-	1	PI
24	3887298E+03	3444433E+00	33346E+00	-	1	PI
25	3887298E+03	3444433E+00	33346E+00	-	1	PI
26	3887298E+03	3444433E+00	33346E+00	-	1	PI
27	3887298E+03	3444433E+00	33346E+00	-	1	PI
28	3887298E+03	3444433E+00	33346E+00	-	1	PI
29	3887298E+03	3444433E+00	33346E+00	-	1	PI
30	3887298E+03	3444433E+00	33346E+00	-	1	PI
31	3887298E+03	3444433E+00	33346E+00	-	1	PI
32	3887298E+03	3444433E+00	33346E+00	-	1	PI
33	3887298E+03	3444433E+00	33346E+00	-	1	PI
34	3887298E+03	3444433E+00	33346E+00	-	1	PI
35	3887298E+03	3444433E+00	33346E+00	-	1	PI
36	3887298E+03	3444433E+00	33346E+00	-	1	PI
37	3887298E+03	3444433E+00	33346E+00	-	1	PI
38	3887298E+03	3444433E+00	33346E+00	-	1	PI

Table 3-1, The rays generated by GRAY suite from the receiver #9 of the 3-layer model. (circle marks those rays appearing in the previous model, black dot indicates the most disturbing events).

4. COMPARISON WITH THE OTHER ALGORITHM

The goal of seismic data processing is to suppress noise and enhance the signal. Random noise is usually reduced by increasing the number of elements such as detectors or sources in the trace mix. Horizontally travelling noise is suppressed by patterns of sources and detectors and by use of frequency and velocity filters. Noise in the reflection path such as ghosts, reverberations and multiples are suppressed by deconvolution or horizontal stacking.

From previous examples, the algorithm of iterative stacking has shown its excellent ability to cancel noise. Now we want to compare its performance with that of other algorithms used in the different categories of noise suppression. We would also like to know whether or not the iterative algorithm is compatible with the other algorithms if they are conducted consecutively.

4.1 Comparison between the Straight, Weighted, and Iterative Stack in Suppressing Random Noise

This comparison is of special interest because straight and weighted stacks are the most common routine

procedures. The modelling method used for this comparison is similar to the one used by Robinson (1970). The model data is a six trace CDP-gather and it is assumed that all traces have already been properly corrected for NMO; the signal consisting of a sequence of evenly-spaced Ricker wavelets. The time extent of windows was short enough for us to ignore the possibility of the attenuation. The signal on each trace differed only in its scale. The noise on each trace was independently generated by convolving a Ricker wavelet with a random series of spikes.

Signal scale could be arbitrarily assigned to each trace and set to 1.0 for the first synthetic trace. S/N energy ratio was computed from the total signal energy and total noise energy throughout the extent of the window. The window time extent (trace length) is 280 samples.

Figures 4-1 to 4-8 show stacking results from different combinations of signal scale and S/N energy ratio. Each figure shows the 6 synthetic CDP-gather, the desired output (true signal), the straight stack, the iterative stack (the 4th-iteration version), the optimum weighted stack, and the iterative stack after a process of optimum weighting.

In figures 4-1 to 4-5, the modelling CDP traces were assigned equal signal scale but varied S/N energy ratio

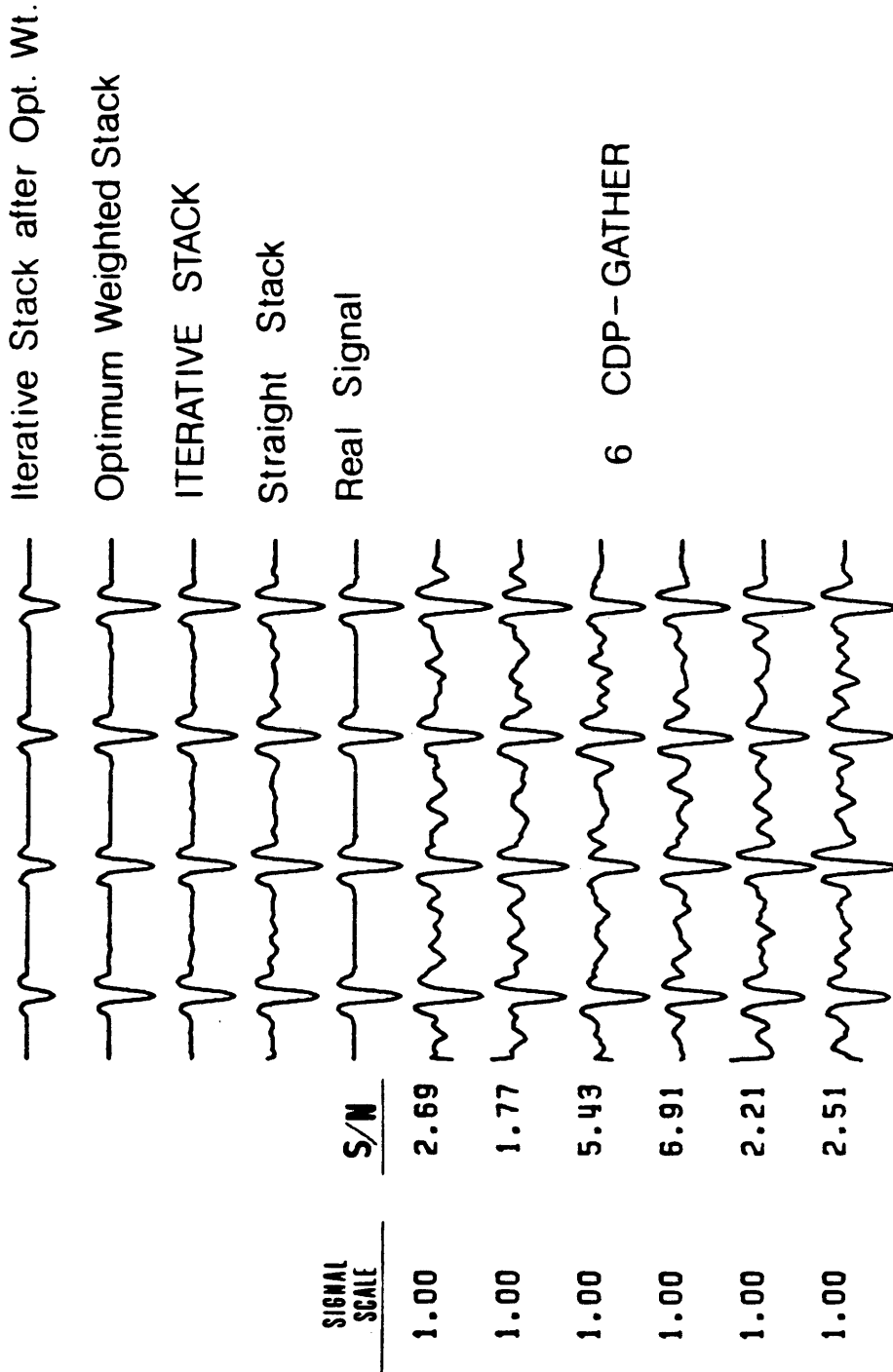


Figure 4-1, Stack from different stacking algorithms with very high S/N ratio.

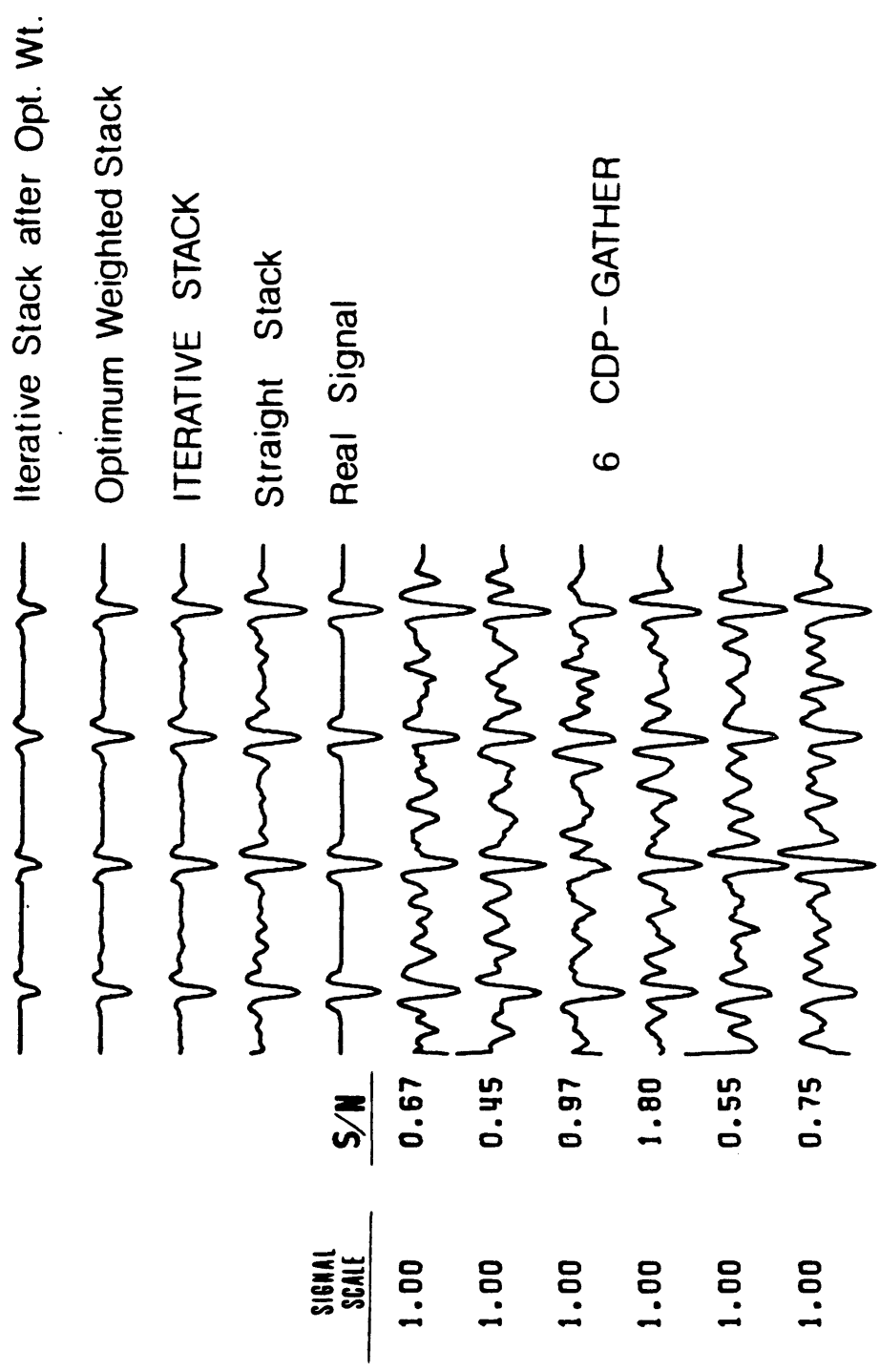


Figure 4-2, Stack from different stacking algorithms with high S/N ratio.

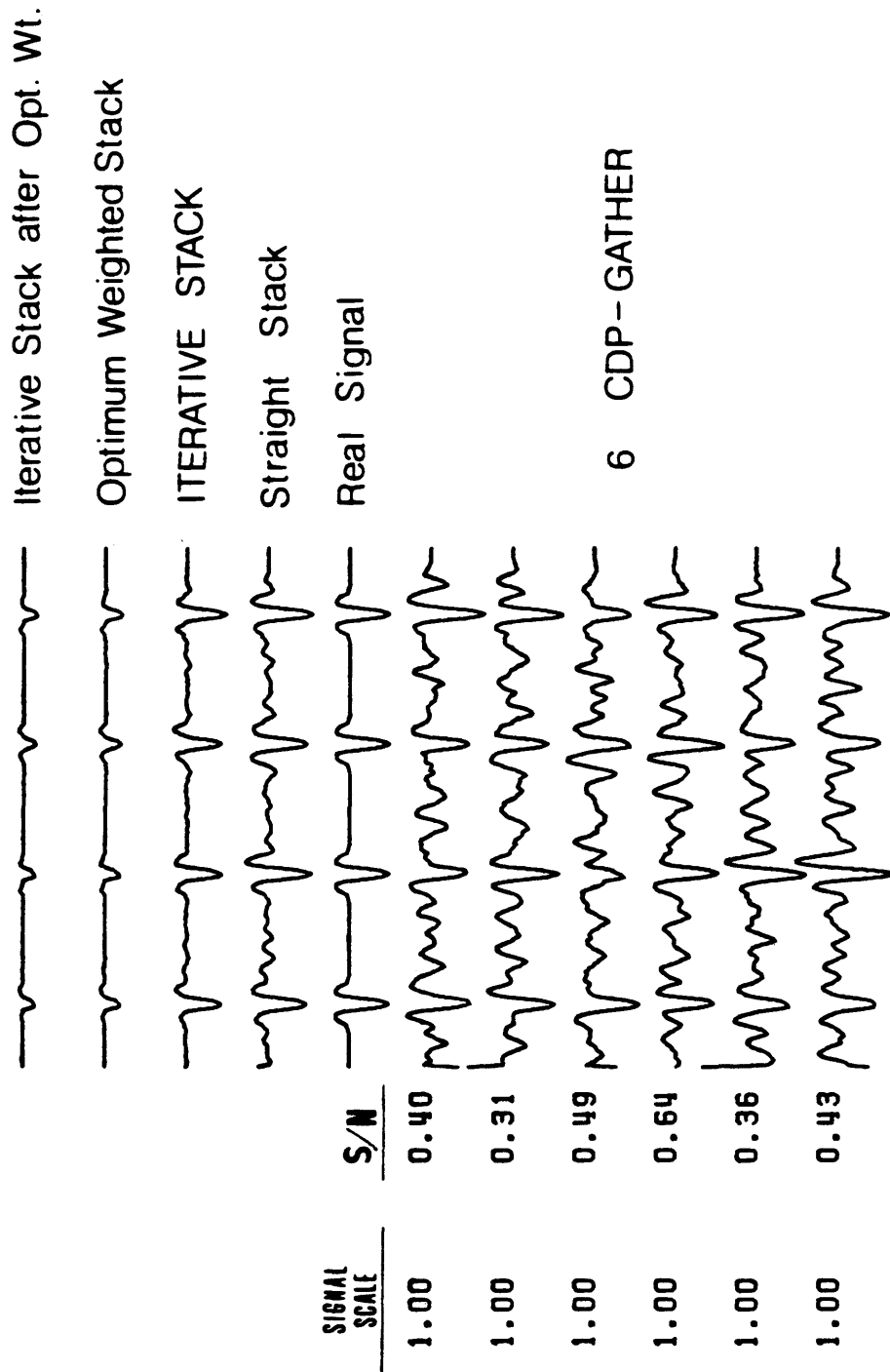


Figure 4-3, Stack from different stacking algorithms with moderate S/N ratio.

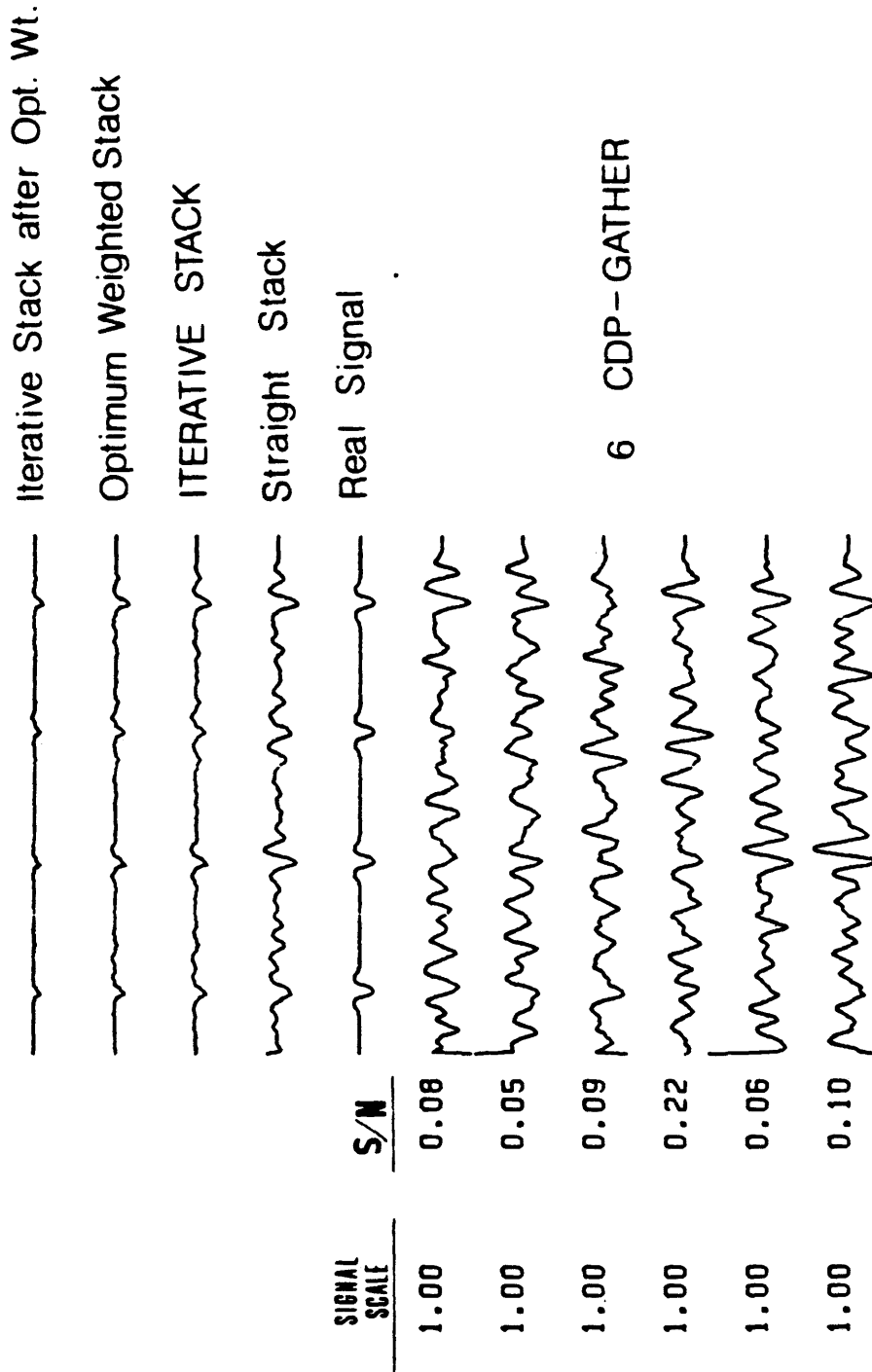


Figure 4-4, Stack from different stacking algorithms with low S/N ratio.

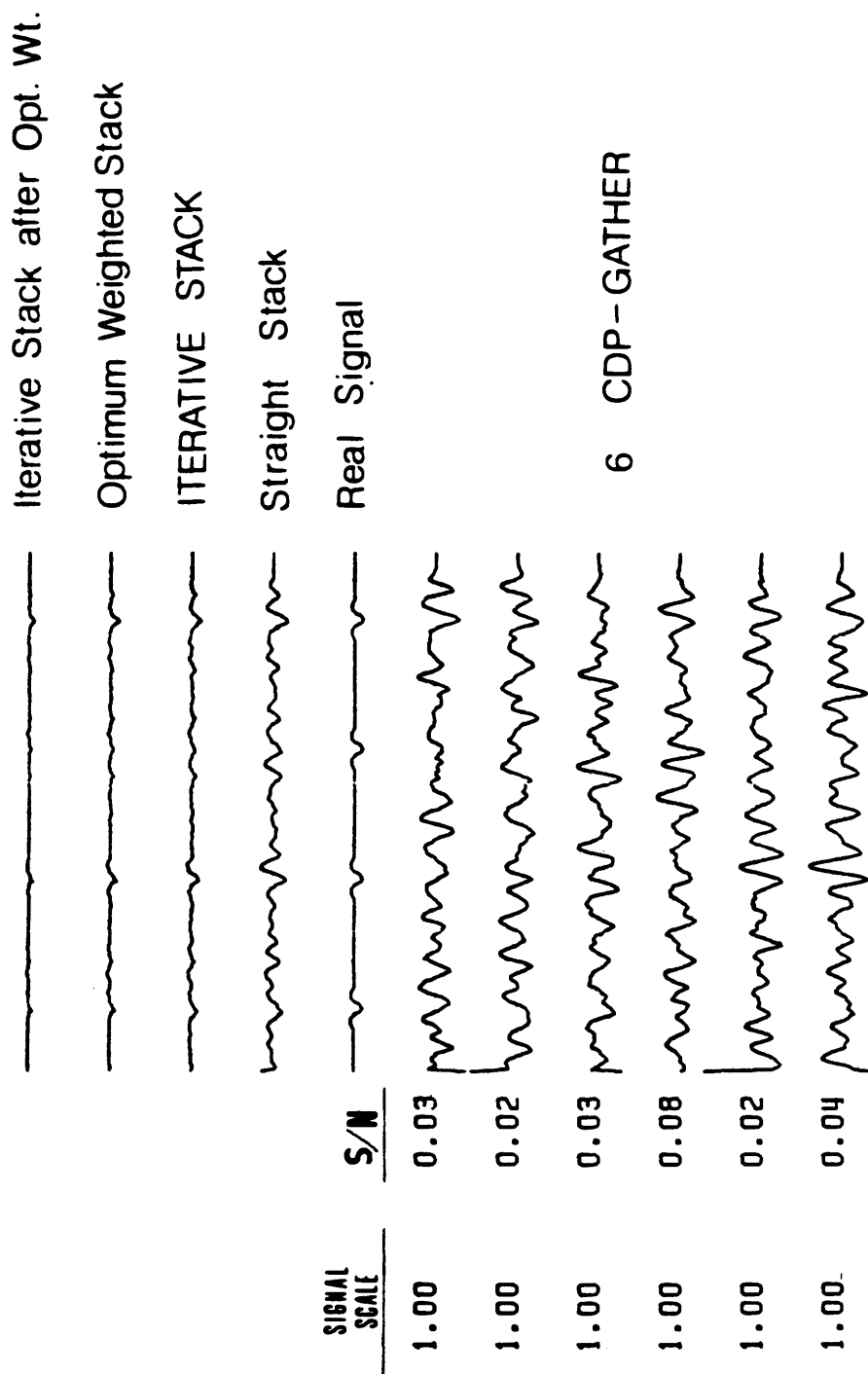


Figure 4-5, Stack from different stacking algorithms with poor S/N ratio.

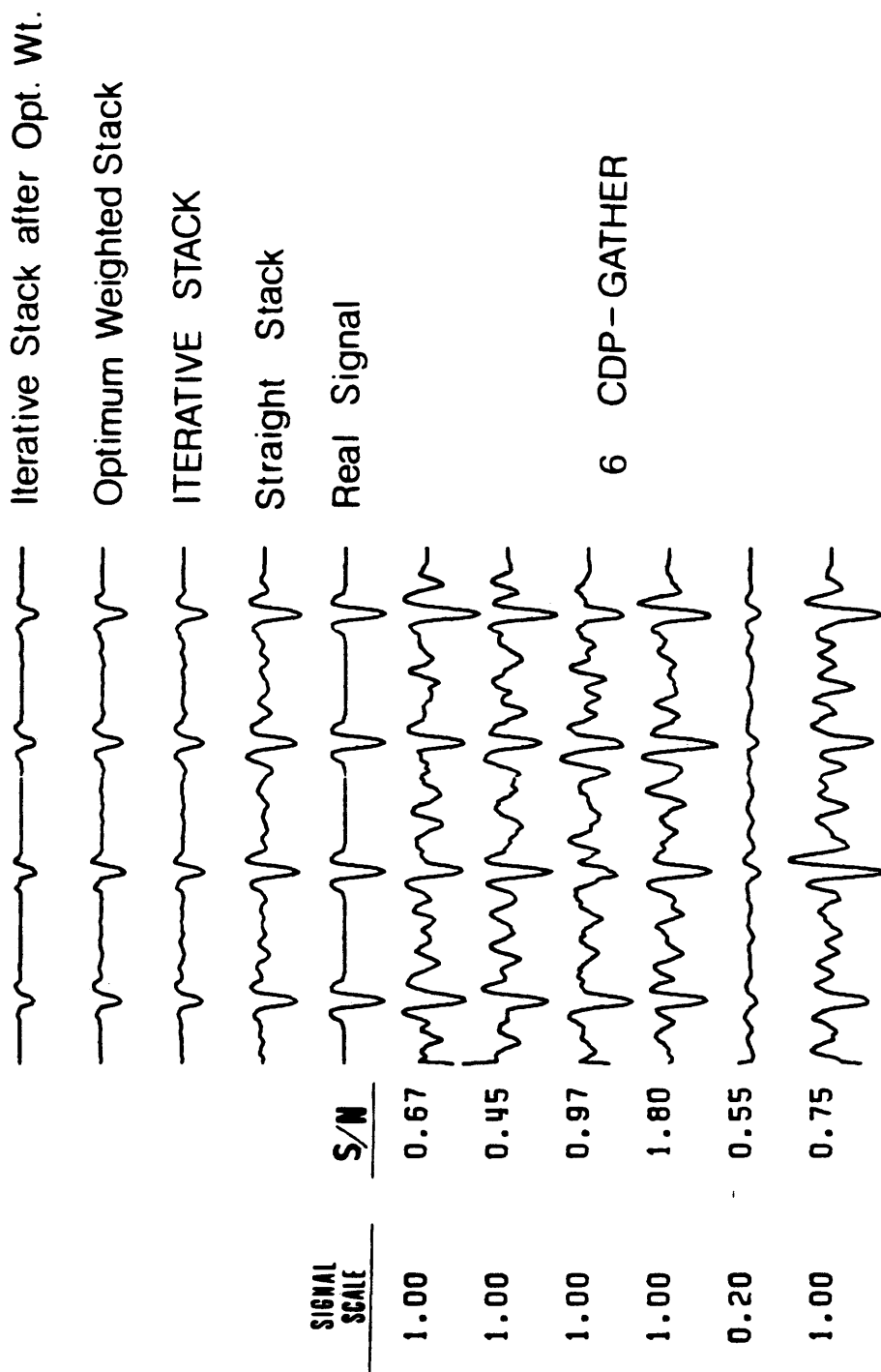


Figure 4-6, Stack from different stacking algorithms with moderate S/N ratio but varied signal scale.

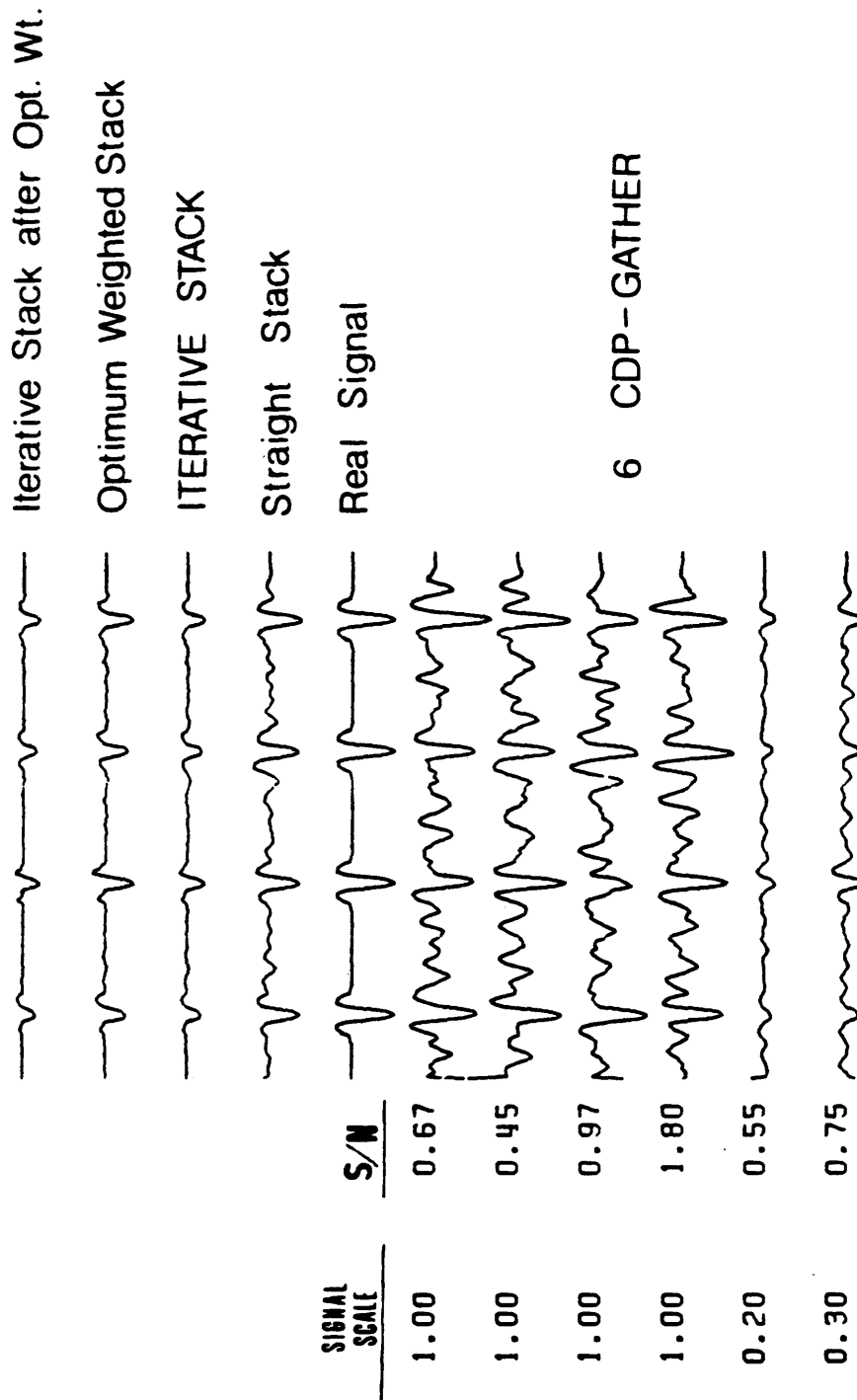


Figure 4-7, Stack from different stacking algorithms with moderate S/N ratio but varied signal scale.

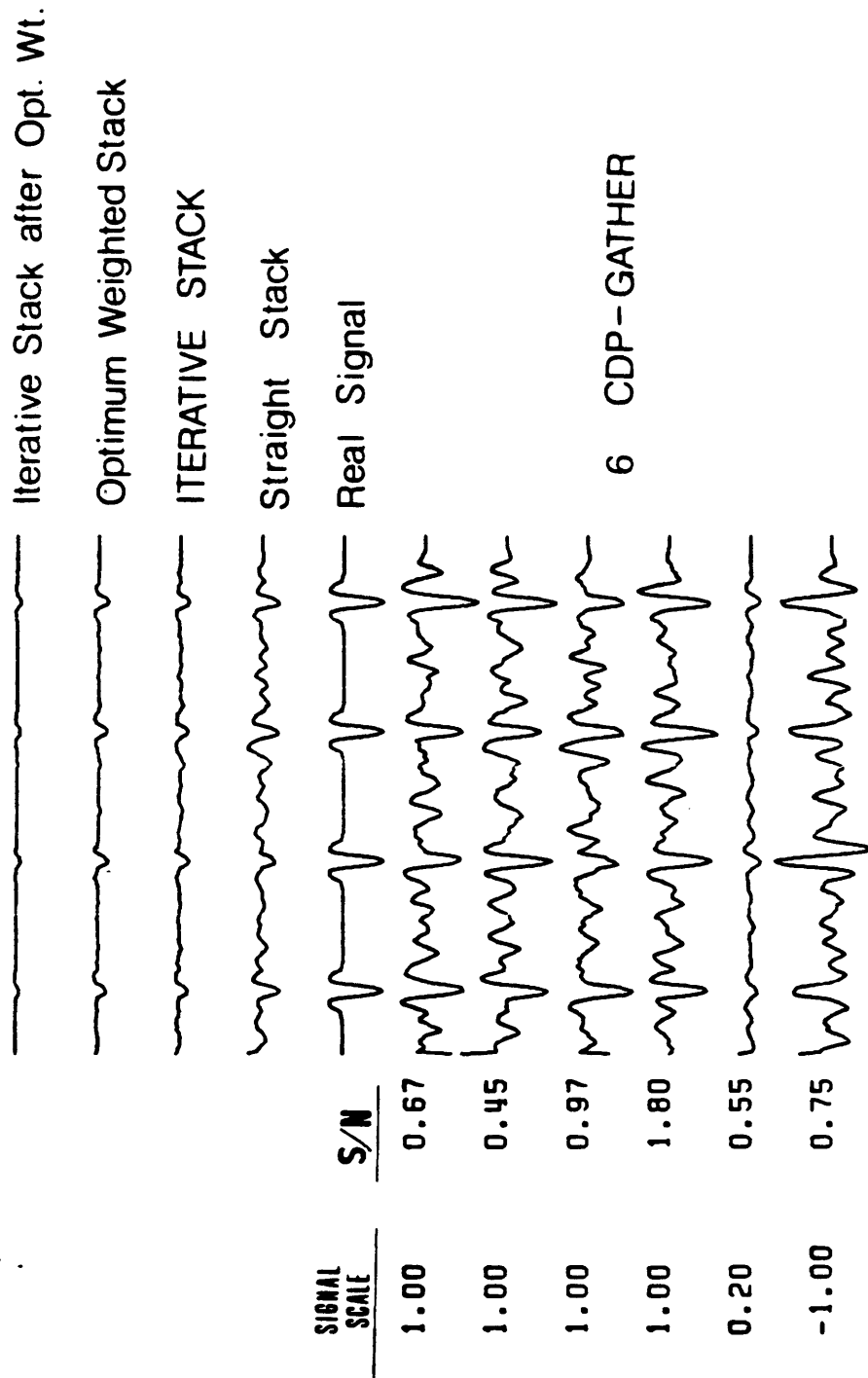


Figure 4-8, Stack from different stacking algorithms with moderate S/N ratio but varied signal scale.

from the highest value (3.5) in figure 4-1 to the lowest value (0.036) in figure 4-5. In figures 4-6 to 4-8, a moderate S/N energy ratio remained constant, while different signal scales were assigned in one or two traces. Therefore, the influence of both the signal scale and S/N ratio on the stacking traces could be observed independently.

The resulting iterative stack is obviously much better than the straight stack in all cases. Hence, the comparison concentrates only on the performance of the optimum weighted stack and iterative stack. My observations could be summarized as follows:

(1) If the signal scale is kept constant among all CDP traces as in the cases in figures 4-1 to 4-5, then:

(1.a) For a high S/N ratio case as in figure 4-1, both optimum weighted stack (OW) and iterative stack (IS) are excellent in enhancing the signal. The iterative stack after optimum weighting (ISOW) shows a smaller output signal than OW or IS.

(1.b) For a moderate S/N ratio as in cases of figures 4-2 and 4-3, IS is more able to emphasize the signal than OW, while ISOW shows no improvement over that of OW or IS and the signal actually gets worse.

(1.c) For a poor S/N ratio as in figures 4-4 and

4-5, IS seems slightly better than OW, and ISOW is still the worst of them.

(2) If the signal scale varied and the S/N ratio were kept about the same as in figures 4-6, 4-7 and 4-8, the result of IS is not so remarkably superior to that of OW as shown in cases (1.b) above, ISOW is improved and might be as good as IS in the case of fig 4-6. The reason is that a weighting procedure tends to pull the signal scales back to an equal level on all traces, which benefits the iterative algorithm of ISOW.

(3) It must be noted that the true optimum weighting stack is based on prior knowledge of the synthetic data in this model. In practice this knowledge is hard to obtain. Robinson (1970) has stated that the best statistical approximation to an optimum stack is essentially one-half as effective as the true optimum technique at improving the S/N ratio on a stack over the ordinary stack. The IS process, on the other hand, does not require any prior assumption about the signal scale or S/N ratio.

(4) The process of OW and IS seems incompatible, which was illustrated by the results of ISOW from those figures. It indicates that the OW process should be avoided in routine seismic work if IS will be executed afterward. However, to make a further analysis, the process of OW

actually consists of two functions: one is to give a big weight to the trace of low signal scale; the other is to give a small weight to the trace of high noise ratio, called the diversity stack. Actually the high noise inside the trace does not bother the performance of IS, while the equalizing of signal scale on all traces will benefit IS. Therefore, the optimum weighting sense for IS is just on the equalization of the signal scale. A cross-product among traces (see equation 6) will provide the value of signal scale of each trace and derive a weight to each trace for scale equalizing. This will improve the performance of IS as in the case of figure 4-7 into the better result of figure 4-3.

In practice, the signal scale for all traces can be assumed to have about the same value. In order to improve the performance of iterative stacking, manual editing of bad traces will still be valuable, whereas diversity stacking should be eliminated from the routine processing sequence.

4.2 Comparison between the Iterative Stack and the Velocity Filter in Suppressing Surface Travelling Noise and Multiples

The objective of this section is to compare the performance of iterative stacking and velocity filtering in suppressing surface travelling noise as well as multiples. Since the velocity filter is commonly used to improve the data continuity prior to the stack, we would also like to know the compatibility of the two processes.

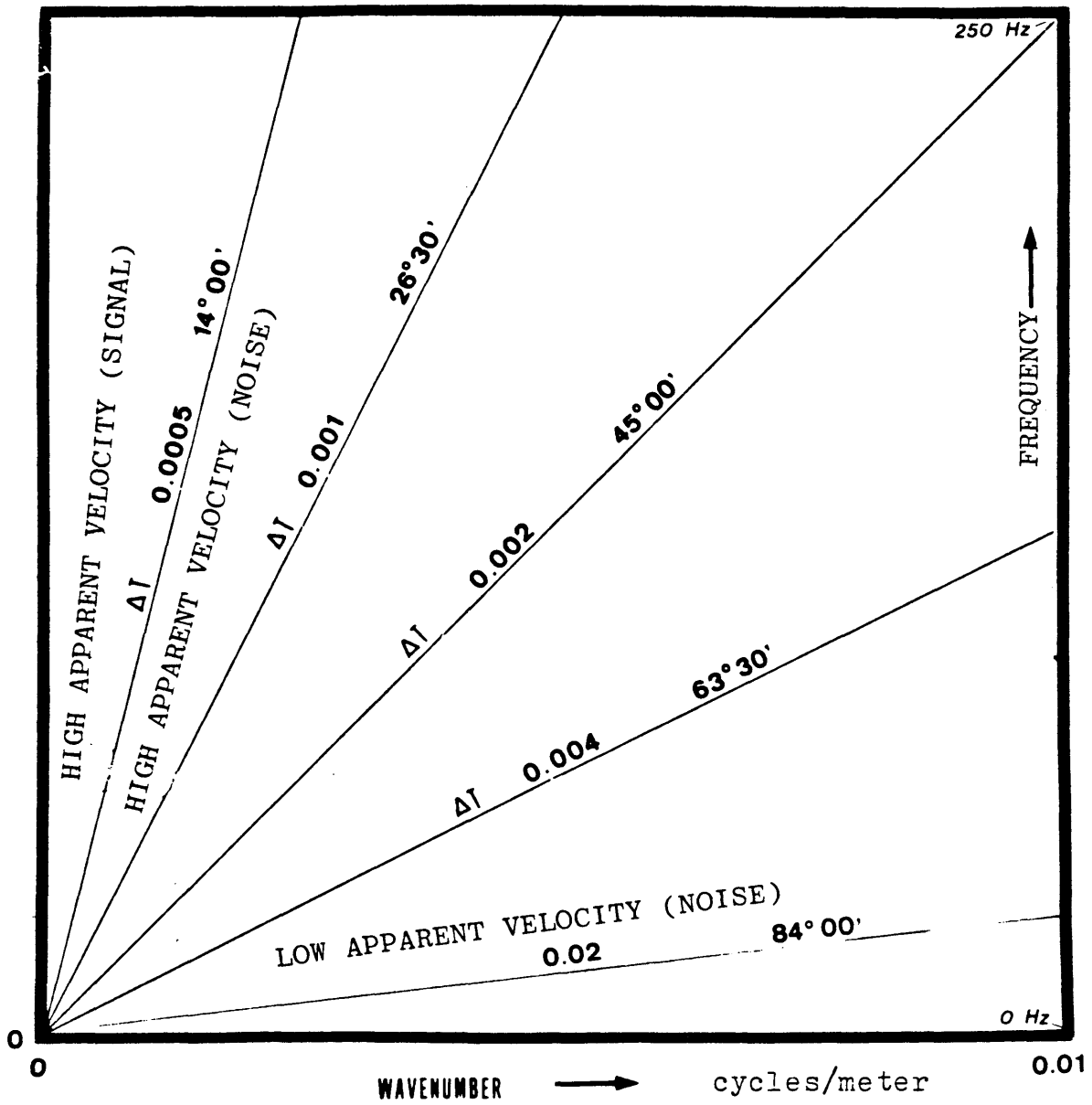
Velocity filters are two dimensional linear operators that can extract signals from a background of coherent noise. These filters may be designed to pass only events with a particular apparent velocity. A typical example of a display of seismic signal and noise spectra in the f-k domain is given in figure 4-9.

Apparent velocity in f-k domain is represented by dip, which is defined as Δt per trace. Noise that consists of responses from surface travelling waves will have a slower apparent velocity; a signal that is reflected at nearly vertical incidence and received by geophones at various distances almost at the same time, will have a much higher apparent velocity (SEG 1974). Therefore, noise which overlaps the signal in x-t space could easily be separated

and filtered in the f-k domain. Also primary reflections and any overlapping multiples usually exhibit different time moveouts which can allow us to selectively discriminate against multiples by using f-k processing on NMO-corrected CDP-gathers (Sengbush 1983). The velocity filter may also be oriented to fit particular dips in the final seismic section to enhance geologic structure (SEG 1974).

The model data used here is generated from the geologic model of figure 3-1. It consists of one waterbottom and two primary reflections and five multiples confined in the water layer (long-period multiples). An extra horizontally travelling event is added for comparative purposes. Figure 4-10 shows the generated CDP-gather after NMO correction. Assuming that NMO corrections have been adequately applied, the primaries are flat. Hence, a velocity filter could easily be designed to eliminate all but the flattest noise components from the CDP-gather.

The f-k filtering is accomplished by zeroing out the part of f-k space in which the multiples and noise reside. We will zero out at least half of the spectrum to suppress the multiples. Figure 4-11 shows the zeroing-out area in f-k domain and figure 4-12 shows the data transformation



STEPOUT / BANDPASS RELATION

.002 SEC. SAMPLING

Figure 4-9, Stepout/bandpass relation in f-k domain for CDP-gather traces with sampling rate of 2 ms.

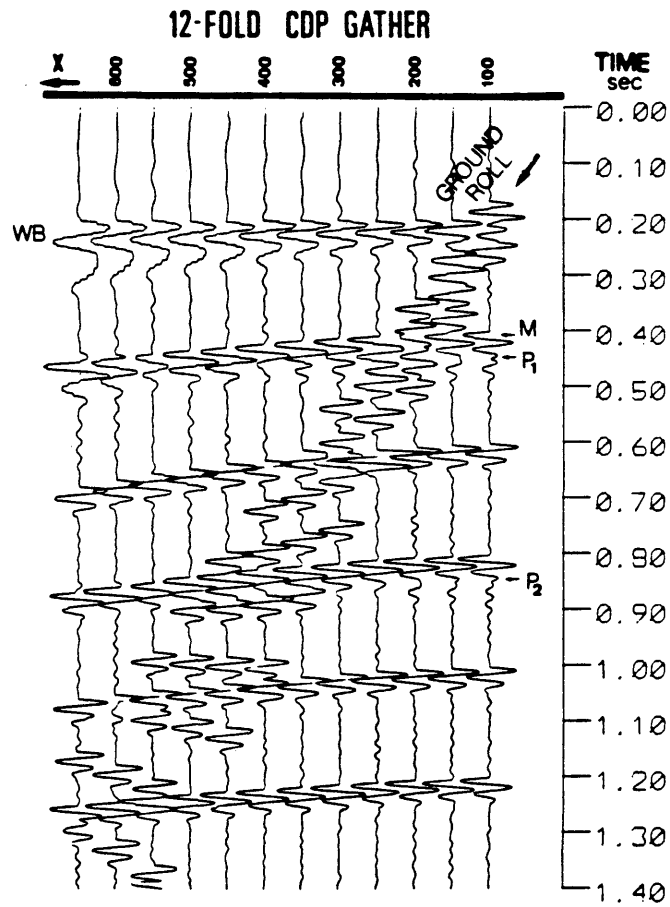


Figure 4-10, Modelling traces generated from figure 3-1 with horizontally travelling noise added, already NMO corrected.

back into x-t space, where surface waves have been successfully eliminated but multiples still exist and seem to get better alignment in the horizontal direction. To compare it with the performance of the iterative stack, we use a CDP-gather of 5-iteration traces from the same model. They are put together in figure 4-12 in order to recognize how significantly the results from these two processings differ.

Another f-k filtering test is made by zeroing below the 45 degree line ($\Delta t=0.002$ sec) in the f-k domain. The masked area includes the surface travelling wave. The resulting x-t plot versus the plot of a simple iterative stack are also illustrated in figure 4-13.

The comparison from these figures show that both methods can delete the surface wave noise, but the iterative stack can totally clear out multiples, while the velocity filter cannot. Even in a model of 36 CDP-gather traces (Sengbush 1983) having better conditions for the f-k filtering process because of more horizontal sampling, was unable to remove multiples in near traces using velocity filtering.

There is no doubt that the iterative stack is more effective than the velocity filter in suppressing low velocity noise, especially for multiples. If that is the

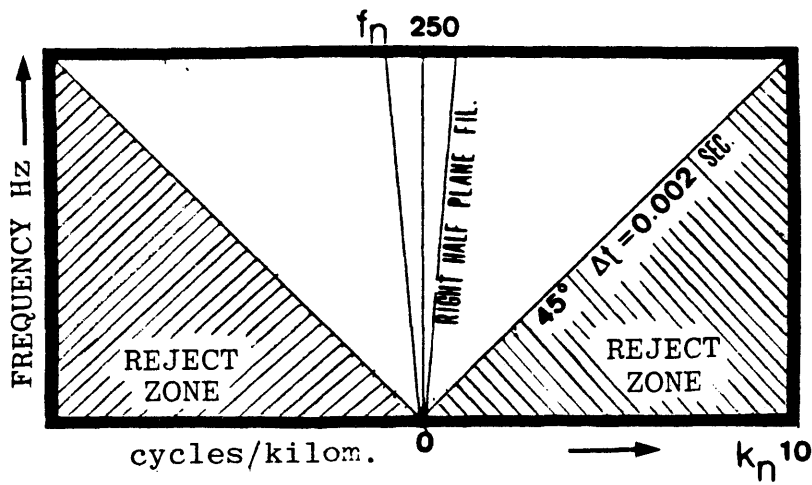
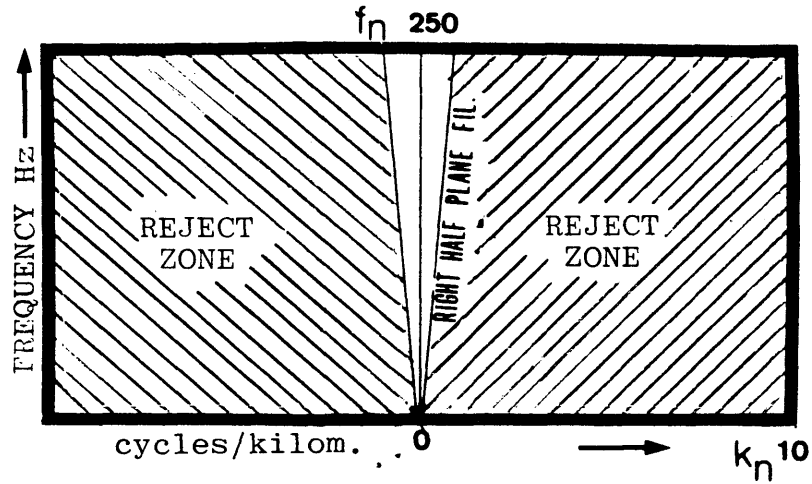


Figure 4-11, Reject zone in f - k domain, upper: Mask for figure 4-12; lower: Mask for figure 4-13.

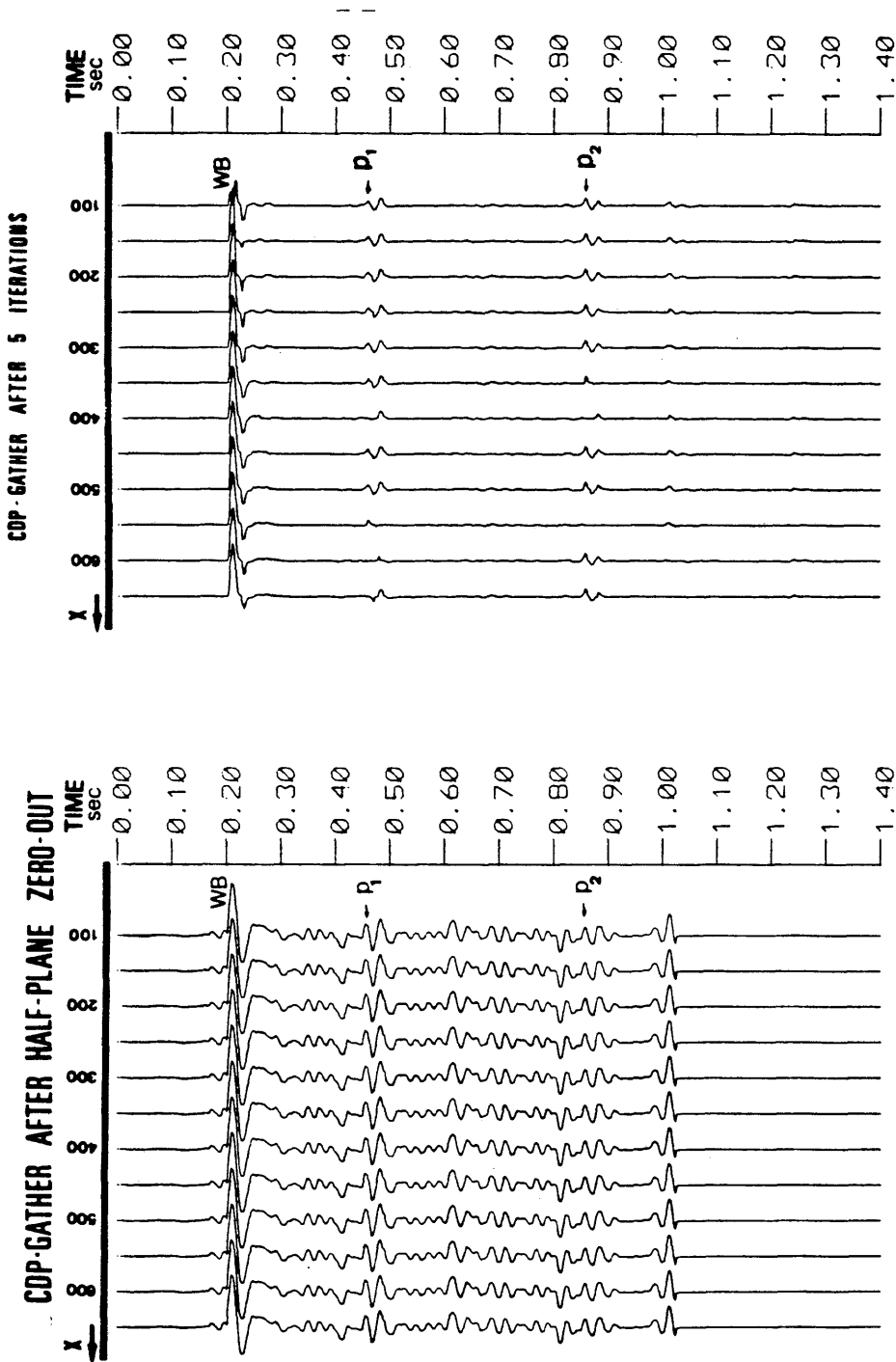


Figure 4-12, Comparison of the f-k filtering result by the half-plane zero-out (left) and the 5-iteration output of the iterative stack (right).

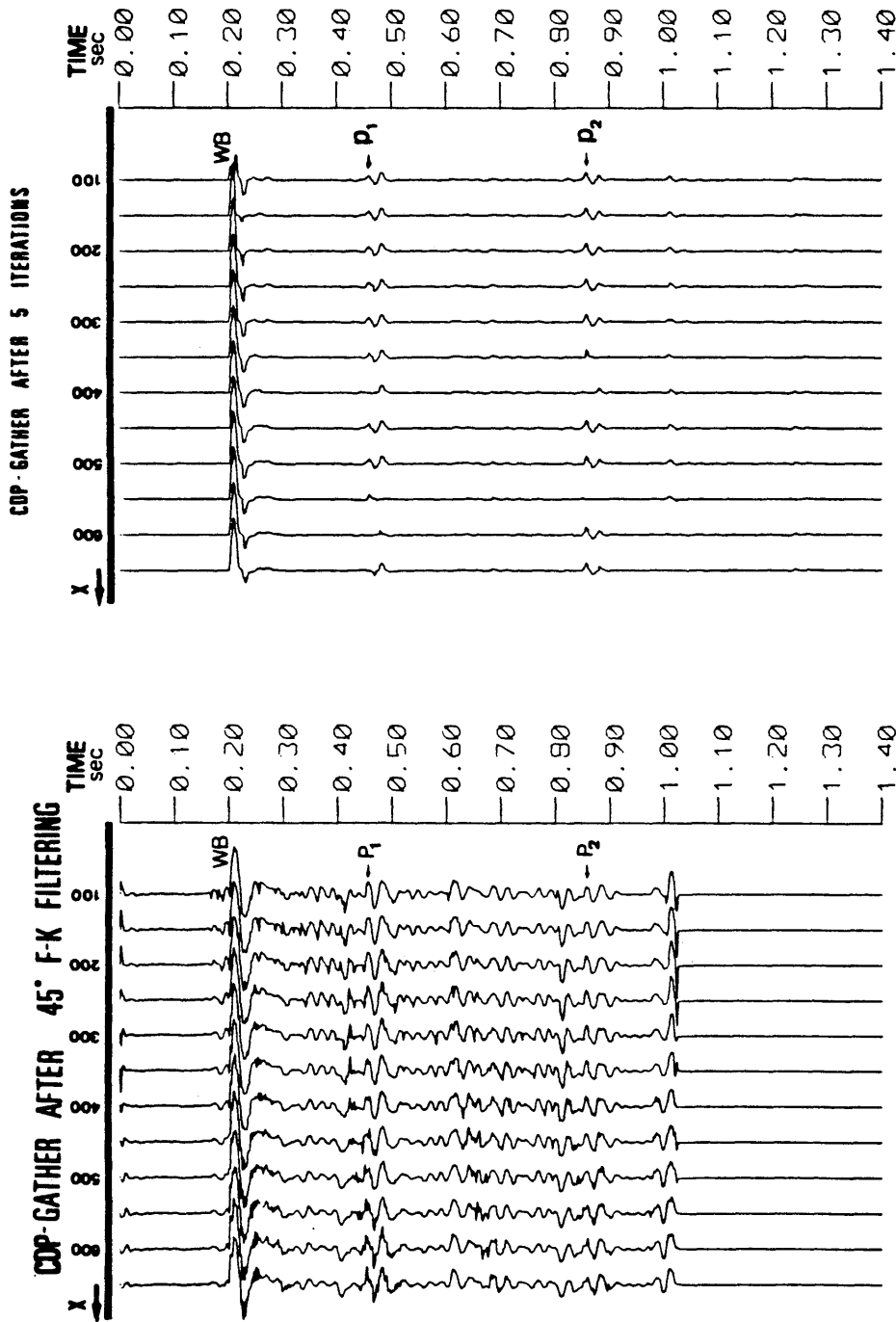


Figure 4-13, Comparison of the f-k filtering result by 45 degree zero-out (left) and the 5-iteration output of the iterative stack.

case, should we eliminate the velocity filtering job in routine seismic work and replace it with iterative stacking? In the conventional routine processing it is usually believed that multiples may be suppressed first in the CDP-gathers by a velocity filter, and then suppressed further by subsequent stacking; if that is true and if iterative stack is subsequently conducted instead of straight stacking, we would like to know if the result is constructive or destructive.

We are reminded that all 2-dimensional filters can be seriously affected by a problem of aliasing in the horizontal direction; signal frequencies which are greater than one half the sampling frequency are folded back into the spectrum as low frequencies equal to the difference between the sampling frequency and the signal frequency. The best solution to avoid aliasing is simply to sample more often horizontally, which means shortening the spacing between receivers. The maximum spacing for proper recording in a given area must be one half of the slowest horizontal velocity divided by the highest signal frequency (SEG 1974).

In our model, the maximum spacing should be less than 20 meters according to that standard, whereas the spacing we used is 50 meters. Thus, aliasing exists and the

frequency content of our signal has been damaged after the 2-dimensional transformations back in the x-t space as in figures 4-12 and 4-13. The iterative stack was applied to these filtered gathers and the result displayed in figure 4-14 (from gather after right-half plane zeroing) and figure 4-15 (from gathers after 45 degree slicing). In the same figures the iterative stack processed directly from original data is also illustrated for comparison. It is evident that iterative stacking after velocity filtering is not able to eliminate multiples as it does individually. Even in higher iterative versions, the multiples still remain unaffected.

Accordingly, we suggest that velocity filtering should be discarded from pre-stacking procedures because its effect is marginal compared to iterative stacking, and it will damage the subsequent iterative stack especially when an aliasing problem exists, which it usually does.

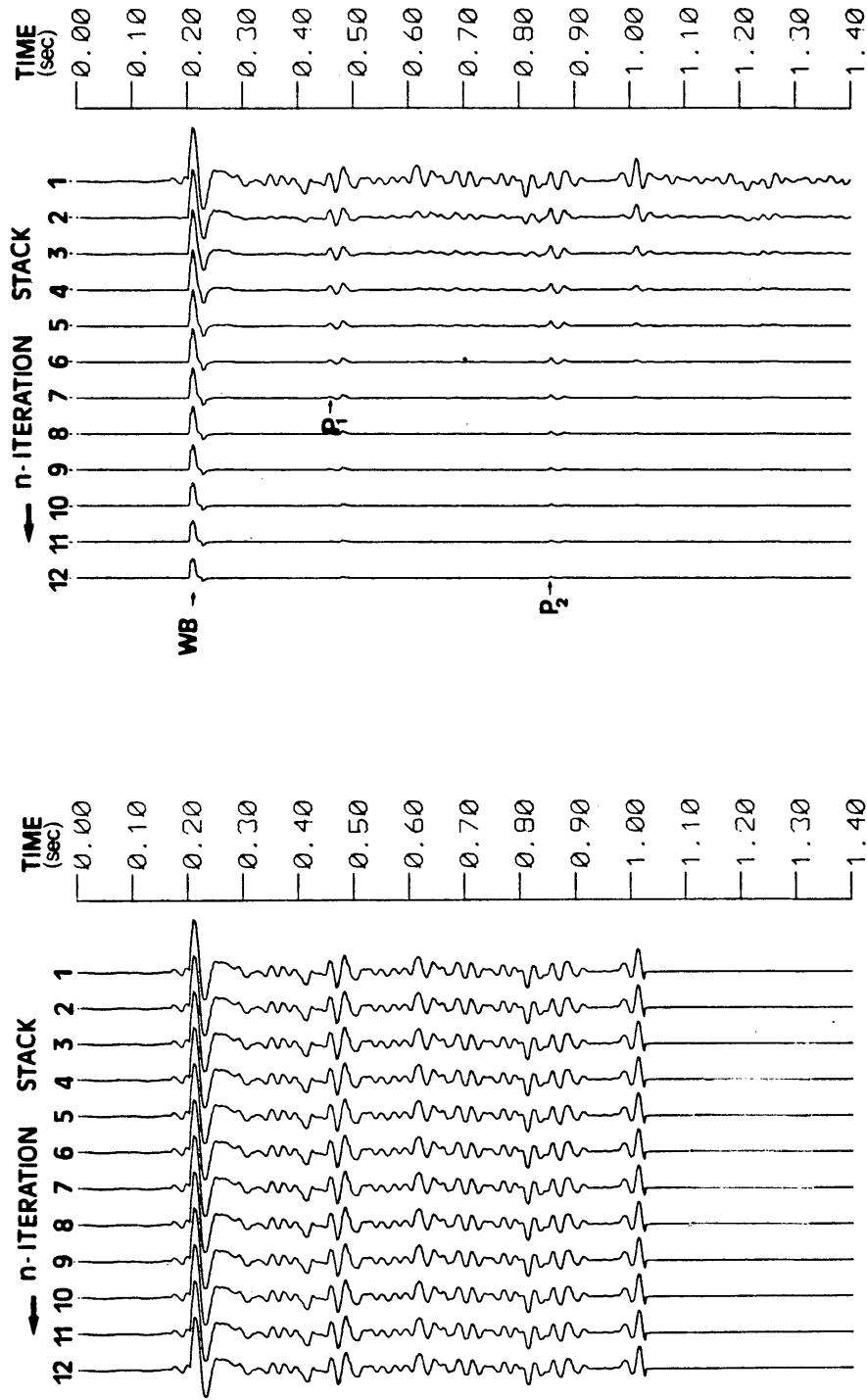


Figure 4-14, Left shows the result of iterative process applied to the output of f-k filtering (half plane zero-out), right picture is the result of direct iterative stack.

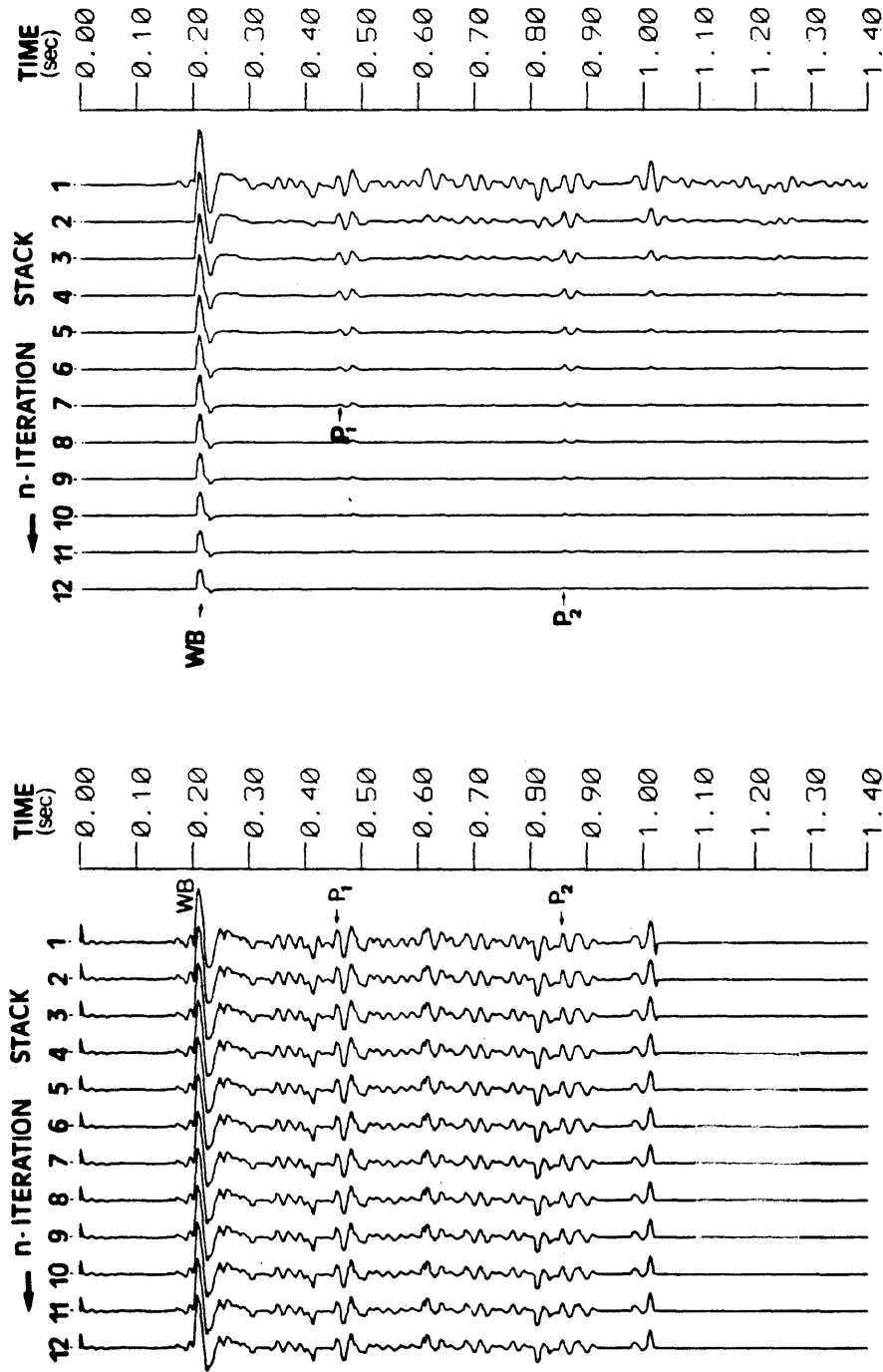


Figure 4-15, Left shows the result of iterative process applied to the output of f-k filtering (45 degree zero-out), right picture is the result of direct iterative stack.

4.3 Comparison between the Iterative Stack and Wiener Deconvolution in Suppressing Periodic Noise

Wiener least-squares inverse filtering (Rice 1962) is one of the most effective tools for the digital reduction of seismic traces, it constitutes the keystone of many current deconvolution methods. The predictive error filtering process, for example, is one of the most flexible deconvolution methods based on that algorithm and, will be used here for our comparison according to the description in papers by Robinson (1966), Robinson & Treitel (1967), and Peacock & Treitel (1969). The Wiener filter is used to deconvolve a reverberating pulse train into an approximation of a zero-delay unit impulse. Generally, it will remove repetitive events having specified periodicities such as ghosts and water-layer reverberations, and occasionally multiples.

The objective of this section is to compare the performance of iterative stack with Wiener deconvolution in suppressing periodic noise. Since deconvolution is usually executed several times in the whole processing procedure before or after the ordinary CDP stacking, we would also like to know the influence of the iterative stack on the Wiener process if iterative stacking is substituted for the

ordinary straight stack.

In order to make a comparative check, we designed a model by adding a ghost reflection to the modelling data of figure 3-1, in which the water layer confined multiples could be regarded as a long-period repetitive event, while the ghost is regarded as the short-period repetitive event. The distortion operator involving the ghost is defined as $S(t)+RS(t-\Delta T)$, where the ghost reflection coefficient R is assumed to be -0.95 and ghost delay time ΔT is 40 ms. Figure 4-16 shows a CDP-gather of 12 traces generated from this model.

Two approaches have been used to attack the ghost and multiple in our modelling traces. One approach is to make the iterative stack first and then apply the Wiener process. The other approach is to do the Wiener process first followed by the iterative stack.

In the first approach, iterative stacking was performed first. Figure 4-17 shows the stacks resulting from the iterative process. Random noise and long-period multiples are totally removed from the final stack after several iterations, but ghost reflections remain in the traces because their velocity is so close to the primaries that their response can not be selectively suppressed. It is important to note that the ghosts of waterbottom layer

have not been exterminated but have been distorted due to the stretching effect of NMO correction. This will distort the subsequent design of the Wiener Operator.

To attack the ghosts in figure 4-17, a Wiener Operator was designed from each trace with a window of 600 samples and filter length of 45 samples. The program to solve the linear equations was obtained from the LINPACK software (Dongara, et. al., 1979), which is in the LIBY: library of the DEC 10 mainframe. The calculated set of filters was convolved with each trace to generate an approximate spike, and then convolved with the 40-Hz weighted sine wavelet to give us the final result as shown in figure 4-18.

The result of the Wiener process shown in figure 4-18 is not satisfactory since the ghost is only slightly attenuated. There are three reasons that probably explain this poor performance. One is that only three reflective events are left in the final stacked traces, which is insufficient to emphasize the periodicity. The other is that ghost responses are distorted by NMO stretching. Third, the wavelet is not minimum phase (example of wavelet spectrum analysis is shown in Appendix A, the program 'TRPLOT').

Since the distorted response of the first reflective

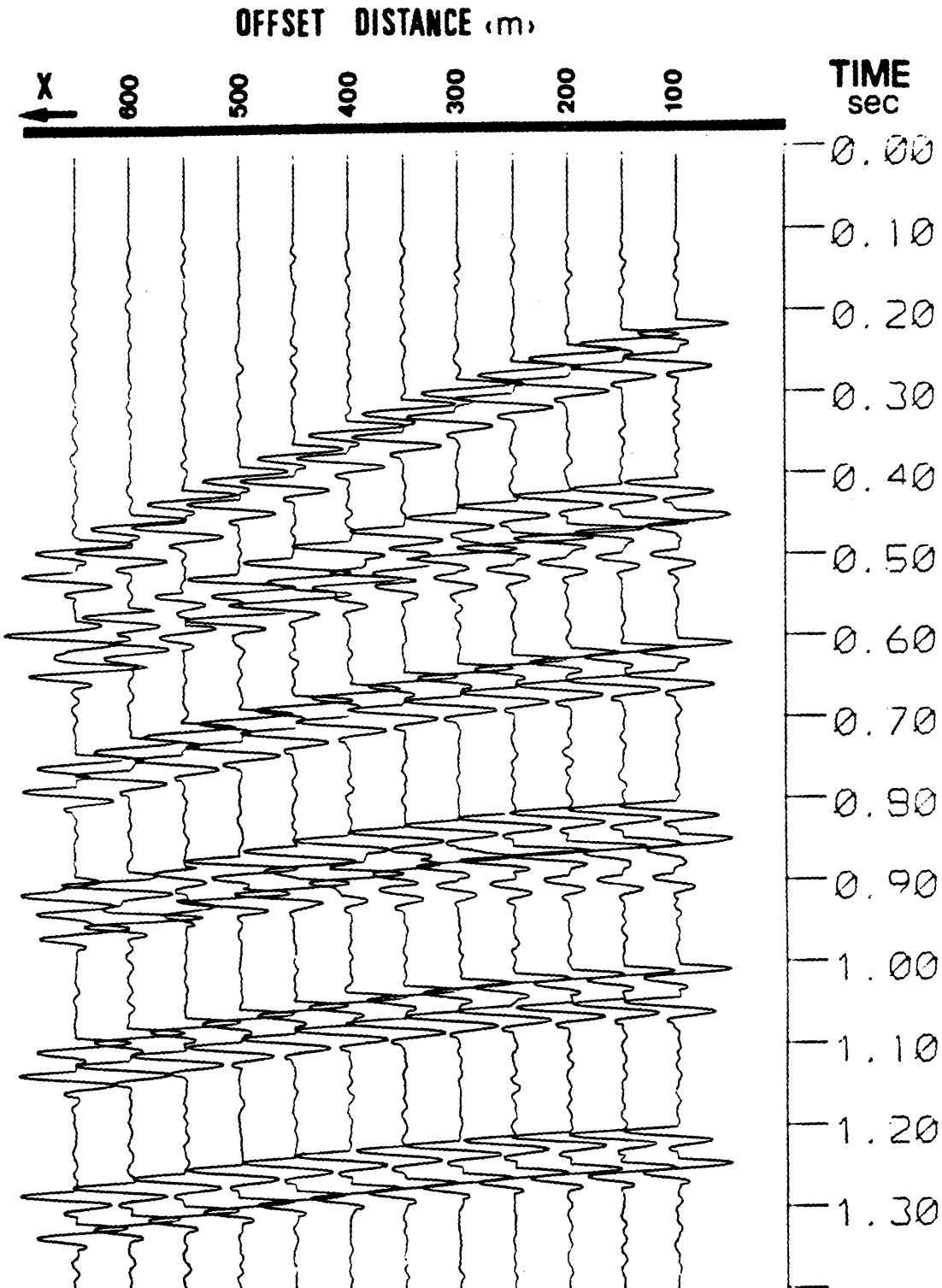


Figure 4-16, 12-fold CDP gather generated from the model in fig. 3-1 with ghost added.

ghost might be the key reason for the poor performance, an attempt was made to develop Wiener filters from the response of the second layer. The result of this Wiener process is illustrated in figure 4-19, which shows some improvement in suppressing ghosts in the second layer but the first layer response was damaged at the same time.

My next attempt to improve the performance of the Wiener process in attacking ghosts on the final iterative stack traces involves the muting skill. Muting is generally performed in the ordinary stacking of CDP-gather. In the early part of the record the long offset traces may be excluded from the stack because they are dominated by refraction arrivals or because their frequency content after NMO correction is appreciably lower than normal traces (Sheriff, 1982) or because the reflection amplitude changes with respect to the incidence angle.

The muting formula I used for that purpose is to exclude data which has NMO time correction over 50 ms. Figure 4-20 shows the resulting stacked traces. Comparing it with the non-muted version of figure 4-17, you will see that the three ghost responses are all sustained in the muted version, and furthermore, the second primary reflection shows greater strength.

The need for the recovery of the waterbottom ghost

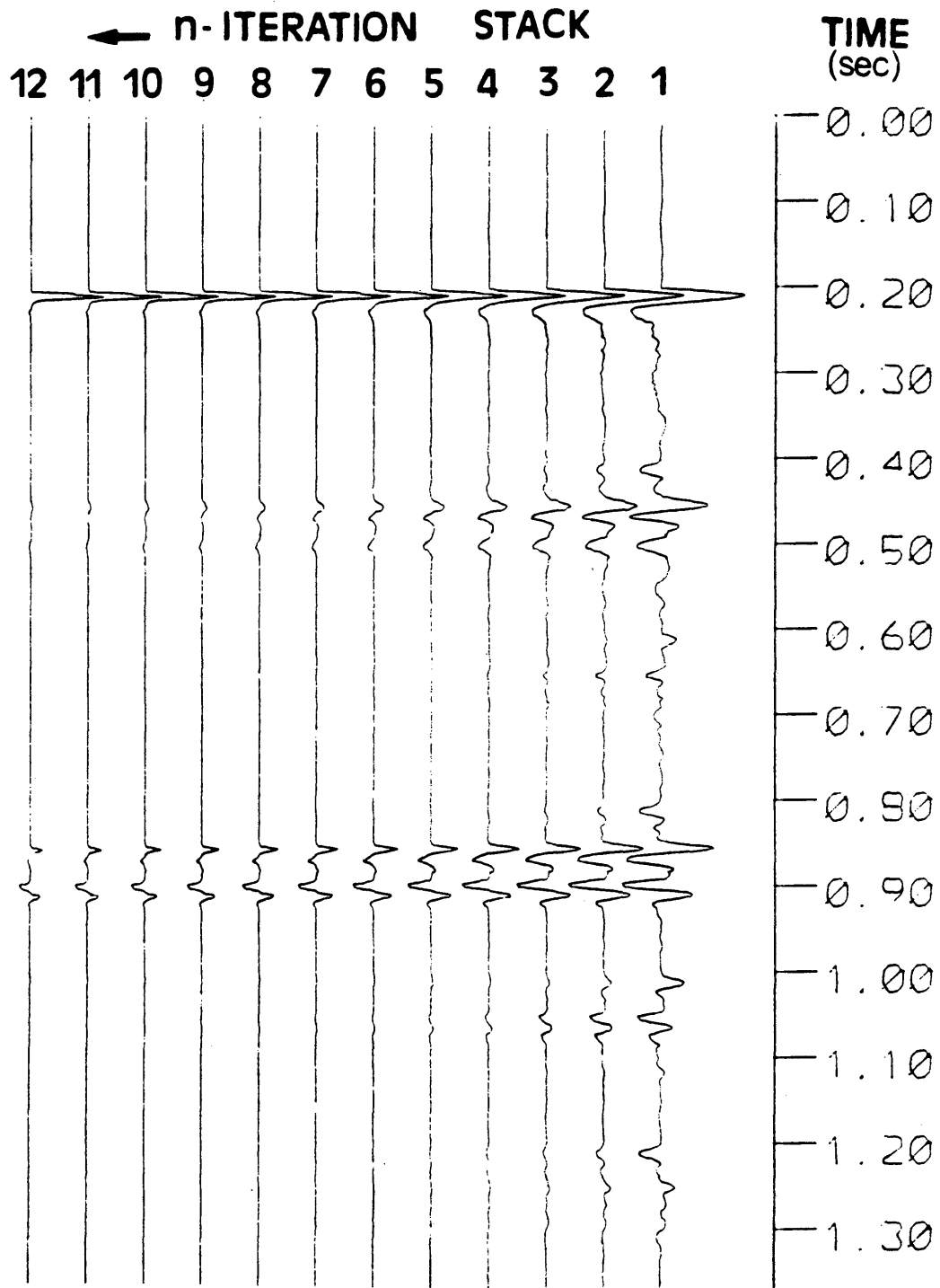


Figure 4-17, Stacks by iterative process on the modelling CDP-gather of figure 4-16.

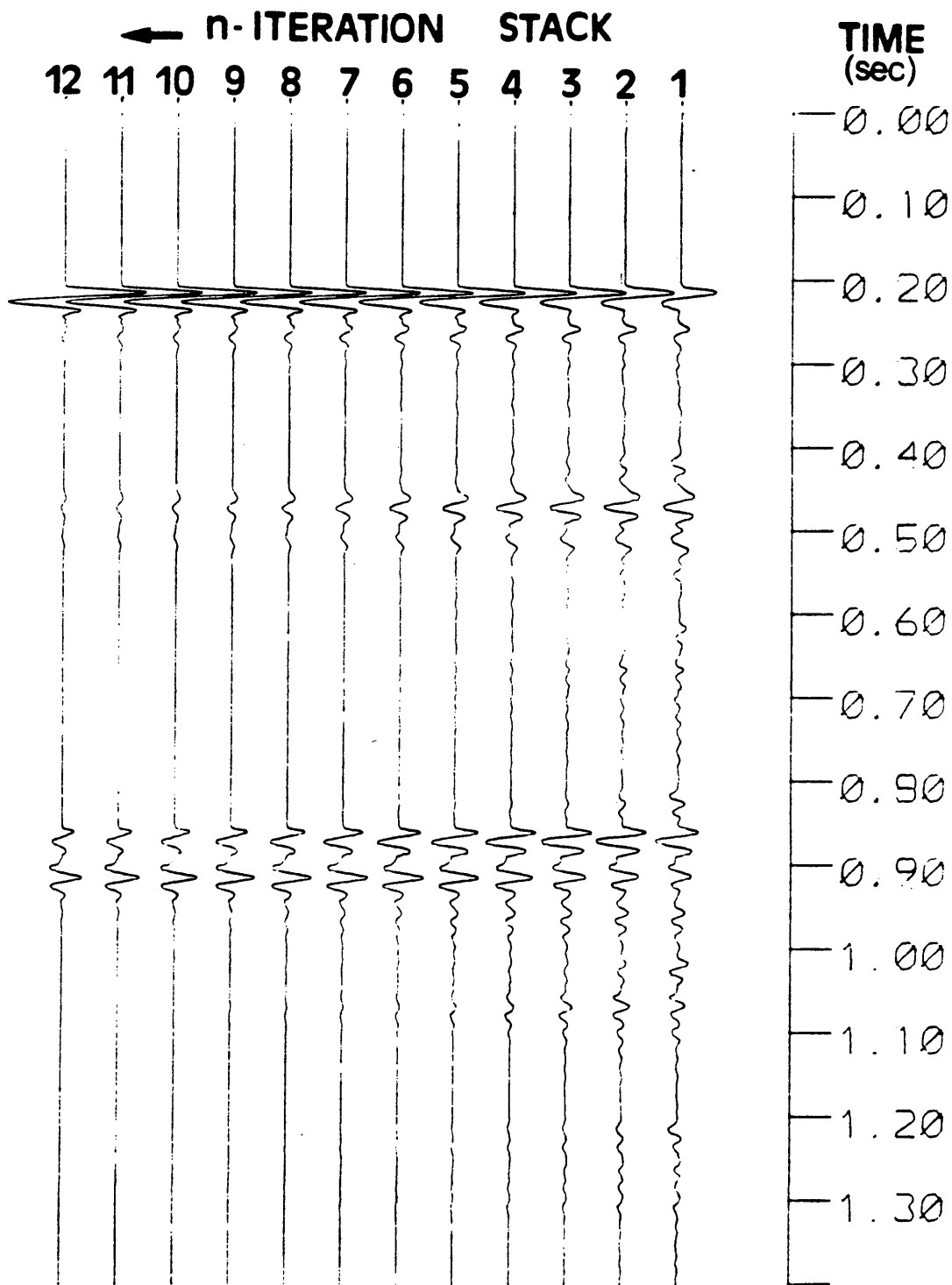


Figure 4-18, The result of Wiener process on figure 4-17.

lies in the fact that this ghost is needed for the design of the deghosting Wiener filter. We performed the Wiener process on this muted version and got the result as shown in figure 4-21. Unfortunately, the ghost, especially at the third layer, was still not suppressed as we had hoped.

It was then guessed that the irregularity in waveform resulting from iterative processing might be the reason that caused the Wiener process to fail, hence, another try was made by using single-trace-processing instead of all-trace-summing for the output of the iterative process, because single-trace-processing has been proven to be better in keeping its original waveform.

Figure 4-22 shows final iterative stacks made by single-trace-processing plus a muting function. Figure 4-23 shows the result after the Wiener process, which is still unable to make any significant improvement in suppressing ghosts. Finally, we tried to use a predictive deconvolution for this job. A predictive error filter was developed from each trace by using an exact ghost-delay time interval of 40 ms. The result of predictive deconvolution is shown in figure 4-24, in which the third layer ghost is still unaffected. If the third layer ghost keeps its original ghost delay, the predictive error process should definitely be effective in removing it.

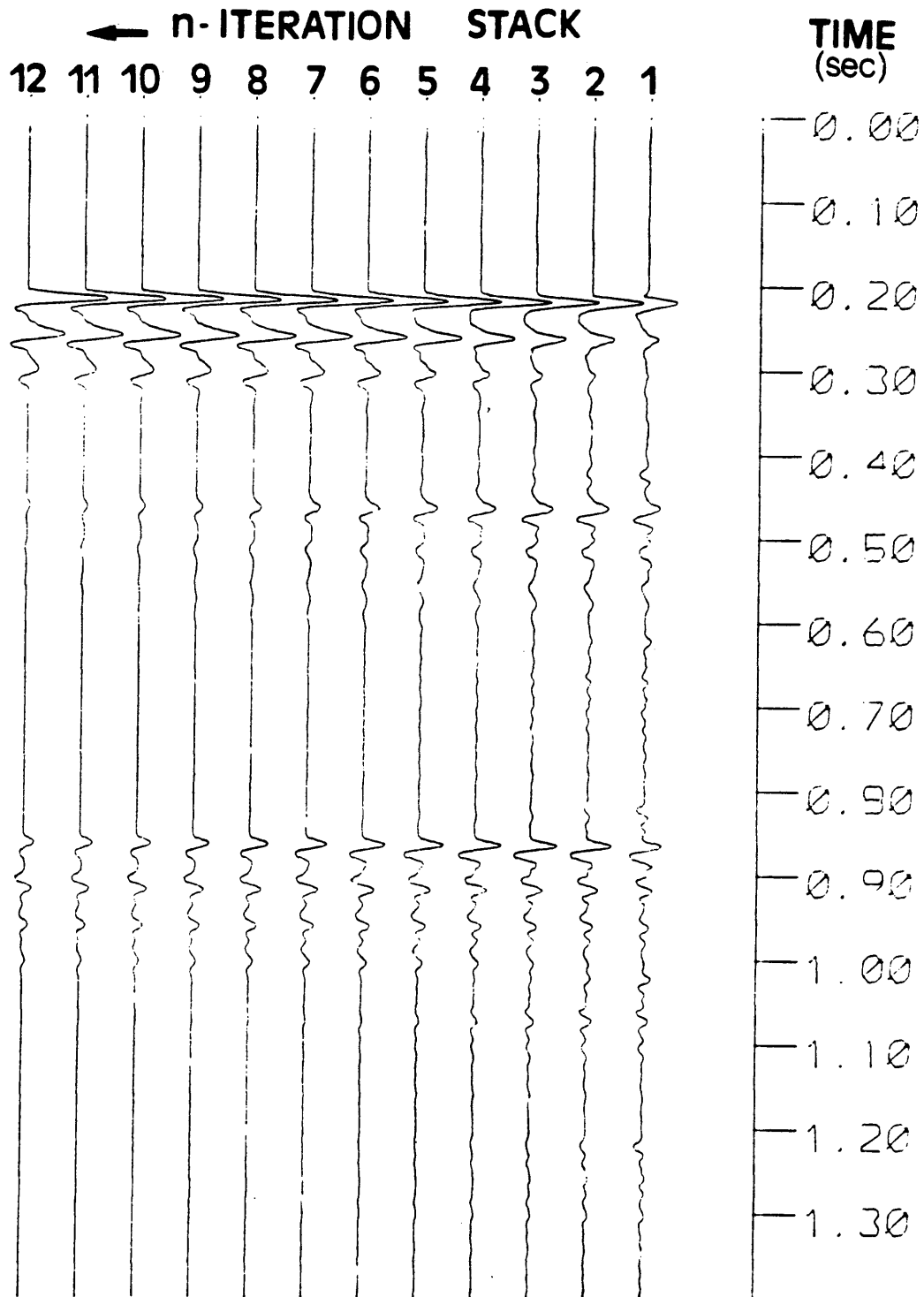


Figure 4-19, The result of Wiener process on figure 4-17 with Wiener operator designed against the 2nd reflection.

Therefore, it is believed that the form of the third layer ghost has been seriously altered after the iterative process because a strong long-period multiple was overlapping it in the original CDP-gather.

If figure 4-24 is the best final stack we can get, we could assume that the normal result of the Wiener deconvolution on the iterative stack traces will be, at most, as good as that shown in the first two layers. However, abnormal results could be seen very often, as happened at the third layer in our test. The conclusion from all these efforts is that the Wiener process could not easily achieve its goal by this approach.

Naess (1979) has stated that "with iterative stack, deconvolution might be better applied between stacks instead of before stack. A few stacks should clean up the traces in the gather considerably and thereby lead to improved design of decon-operators". Our modelling tests just disproved this optimistic assumption.

The second approach, on the other hand, which executes the Wiener process on the CDP-gather traces prior to stacking, has no such problem. First we use the same design parameters to develop the Wiener operator for each original trace, and convolve the trace with the filter to get a de-ghosted trace as shown on figure 4-25. We then

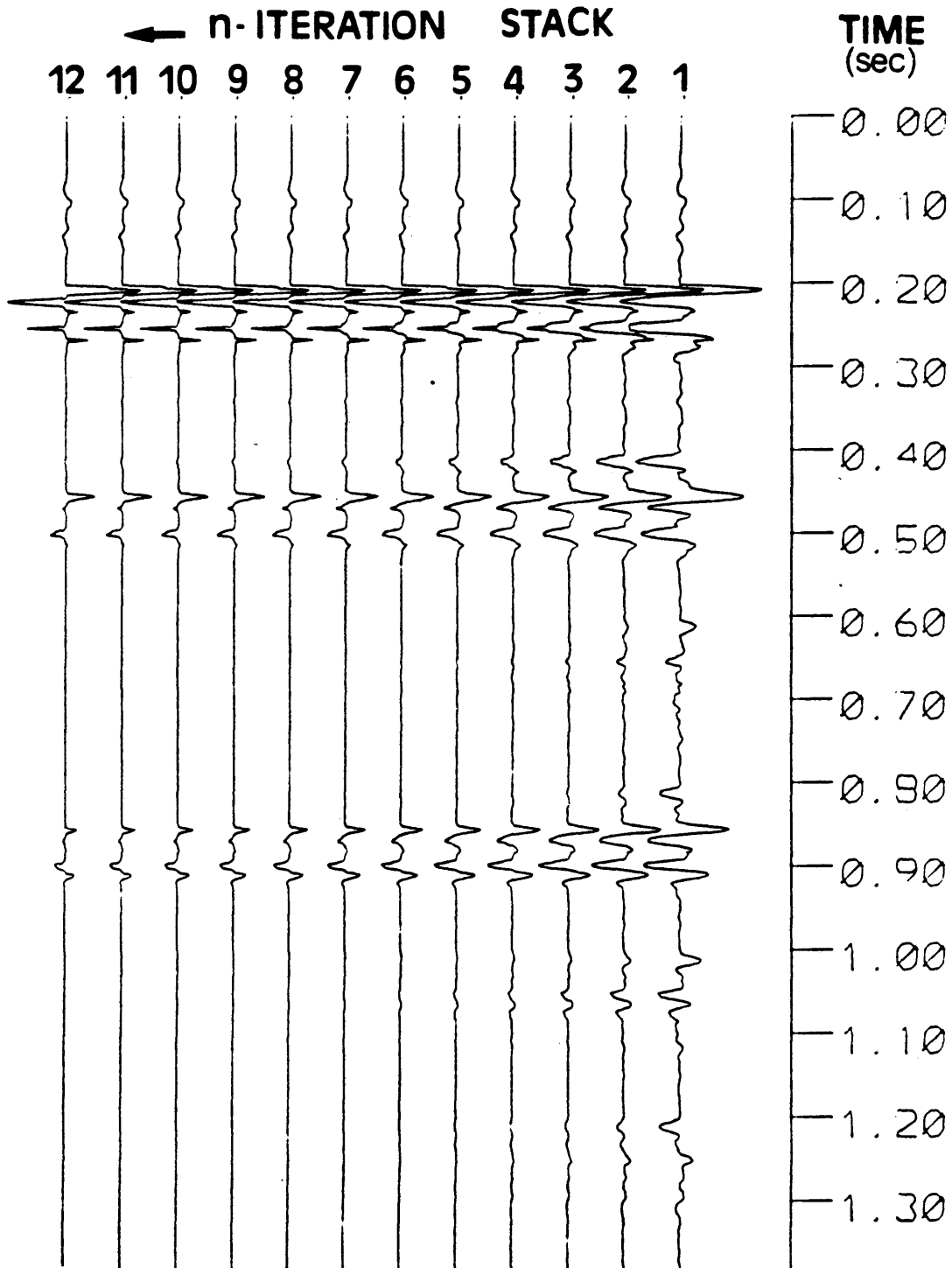


Figure 4-20, Iteration after muting.

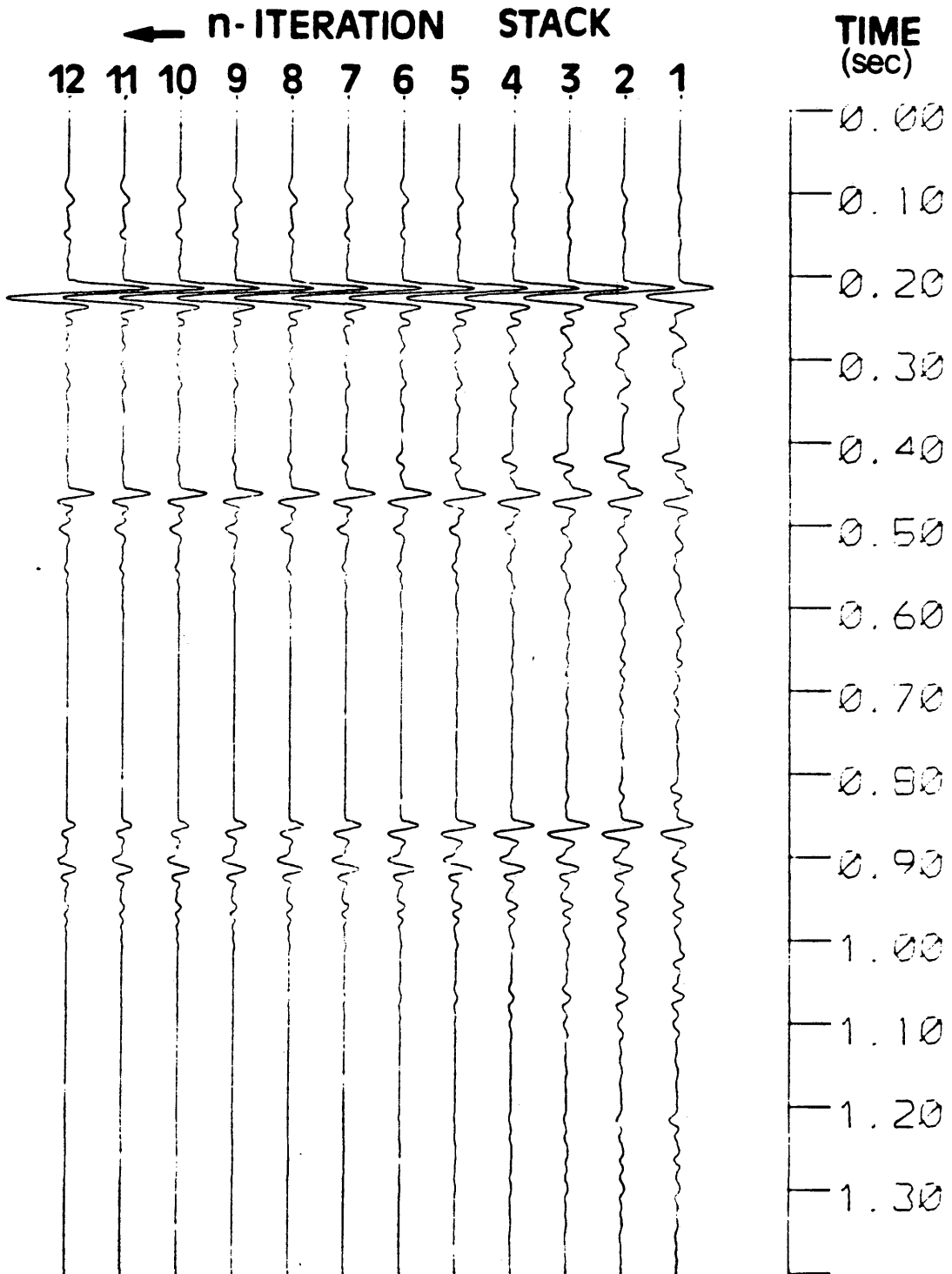


Figure 4-21, The result of Wiener process on figure 4-20.

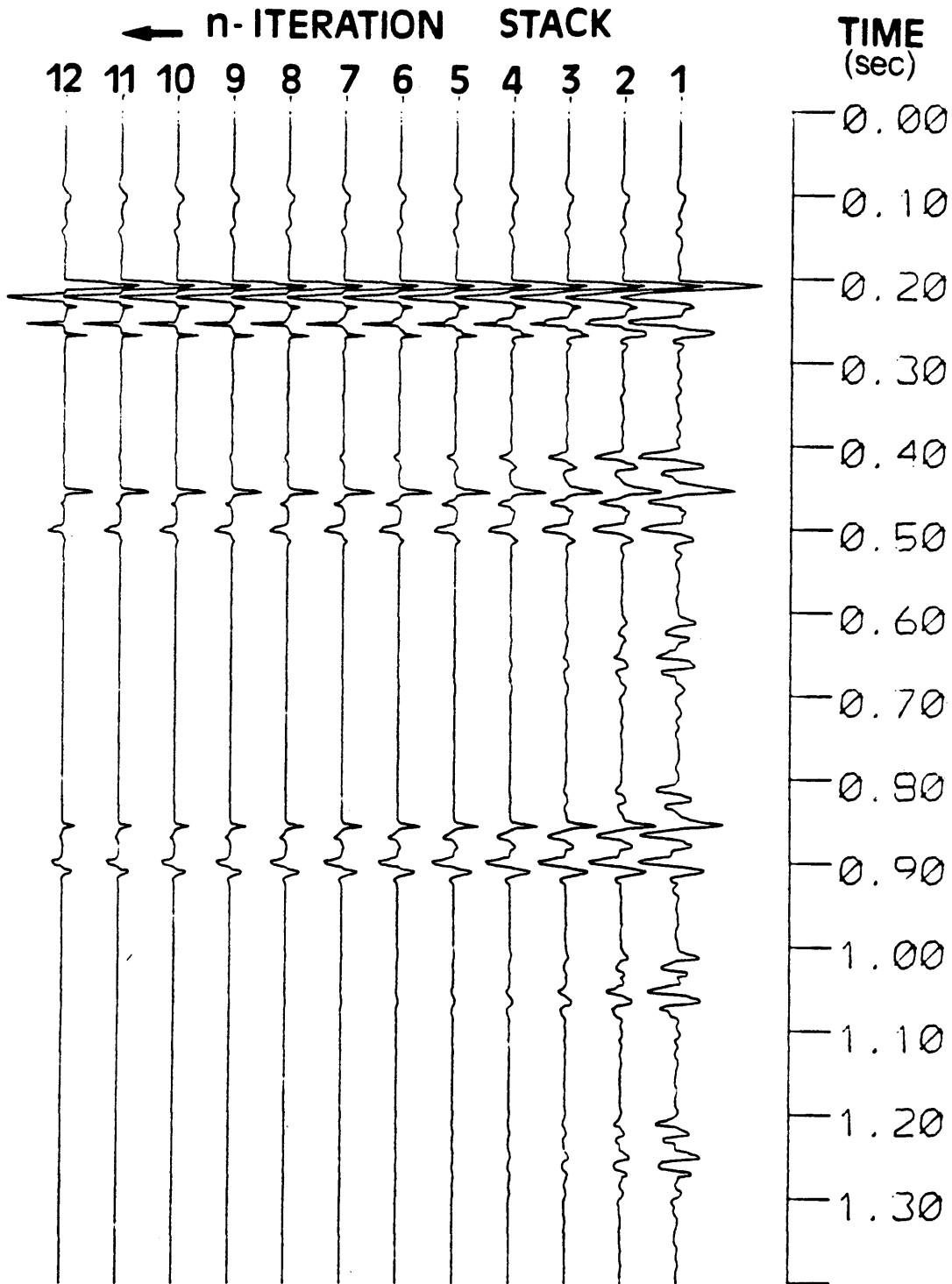


Figure 4-22, The final iterative stacks made by single-trace-processing plus a muting function.

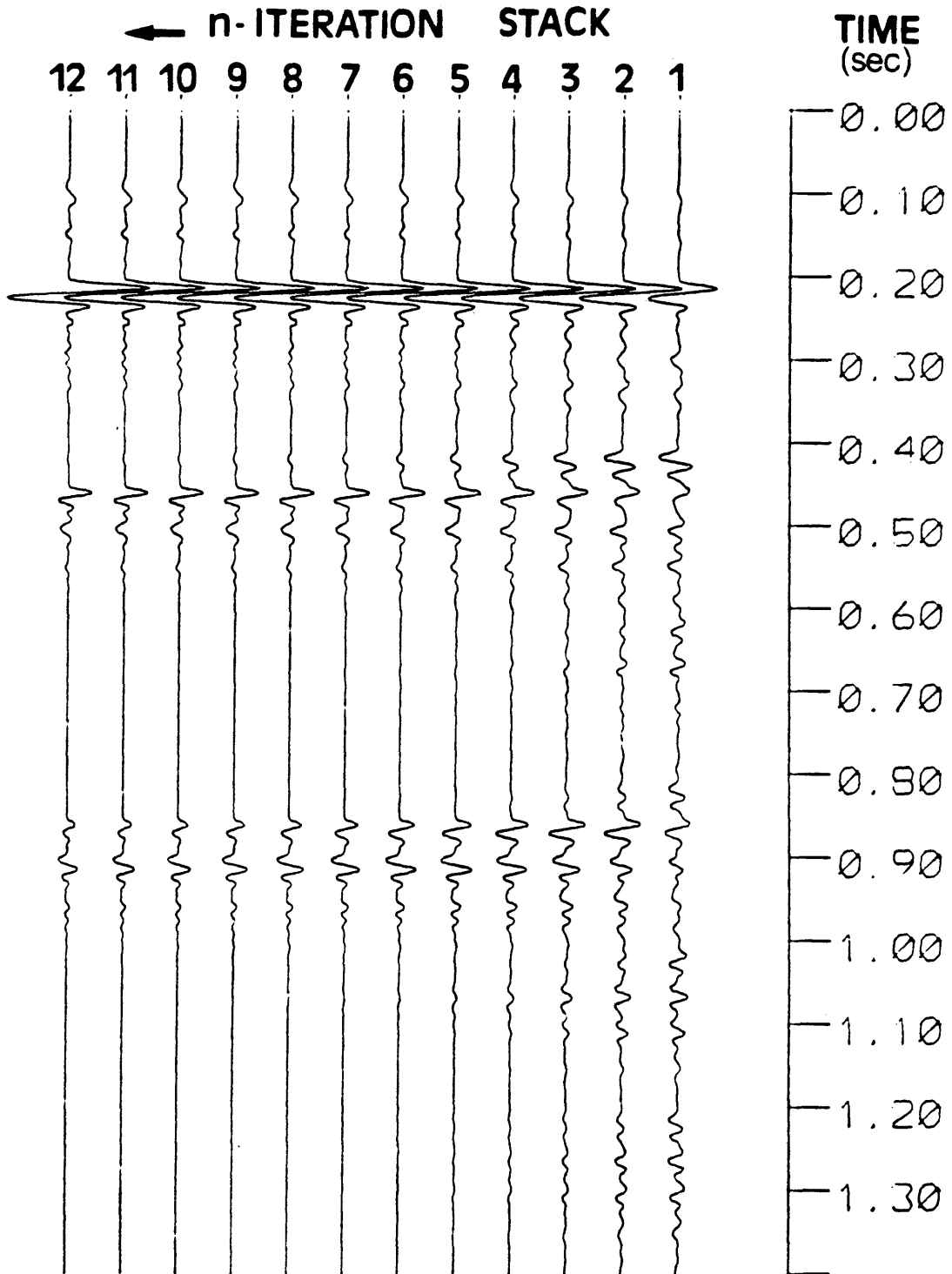


Figure 4-23, The result of Wiener process on fig. 4-22.

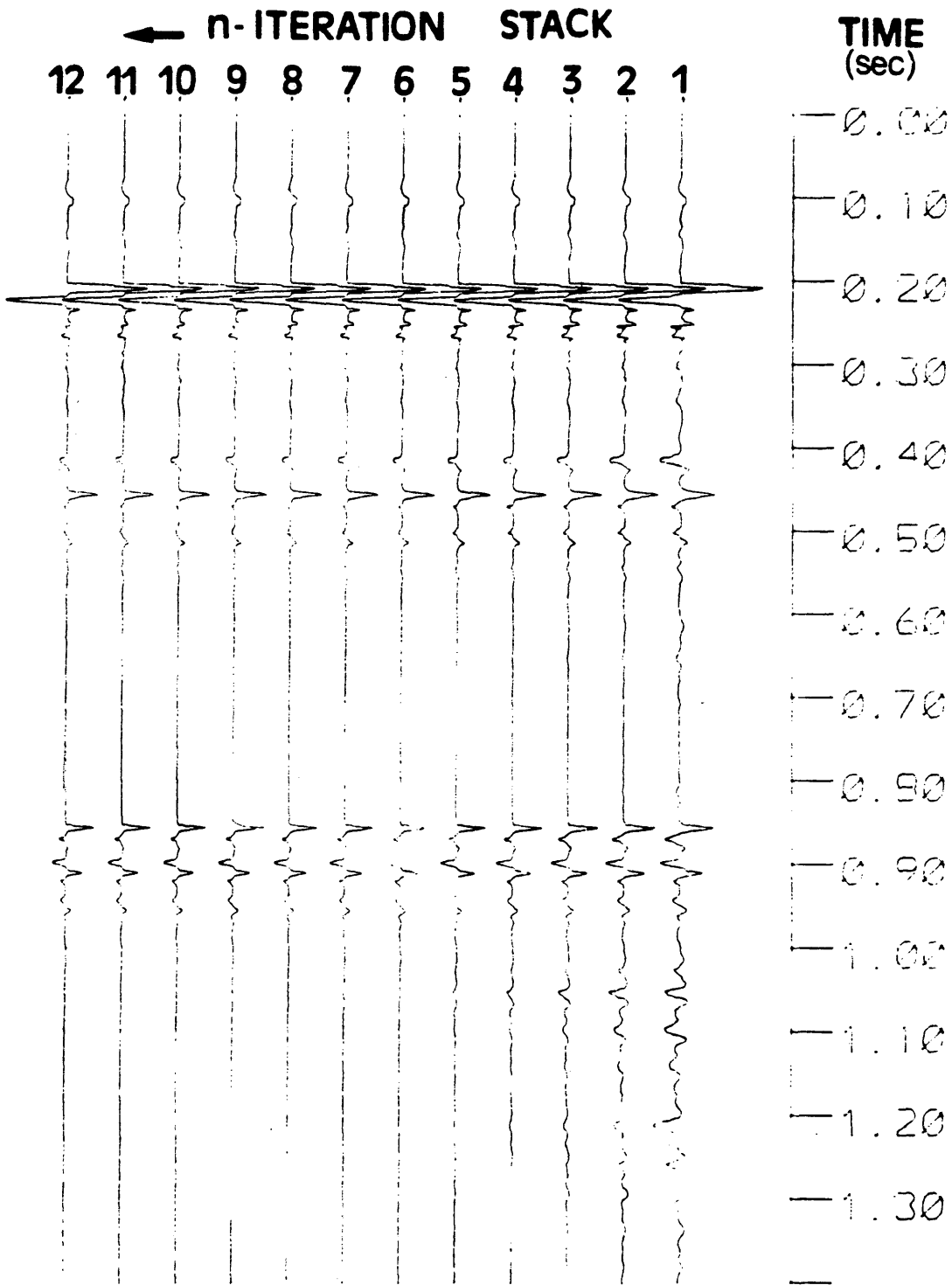


Figure 4-24, The result of predictive deconv on fig. 4-22.

perform an iterative stacking to get a final stack as shown in figure 4-26, where the long-period multiples are deleted in this step. Figure 4-26 shows an almost perfect final section where both ghosts and multiples are almost totally suppressed.

It can be concluded that the iterative stacking (non-linear process) will prevent the subsequent Wiener process (linear process) from giving a normal performance, but, if we apply the Wiener process first it will not bother the subsequent iterative stacking. Thus, in routine procedure we suggest that the Wiener process should be executed prior to iterative stacking. This implies that in general linear processes should, when possible, precede the non-linear processes.

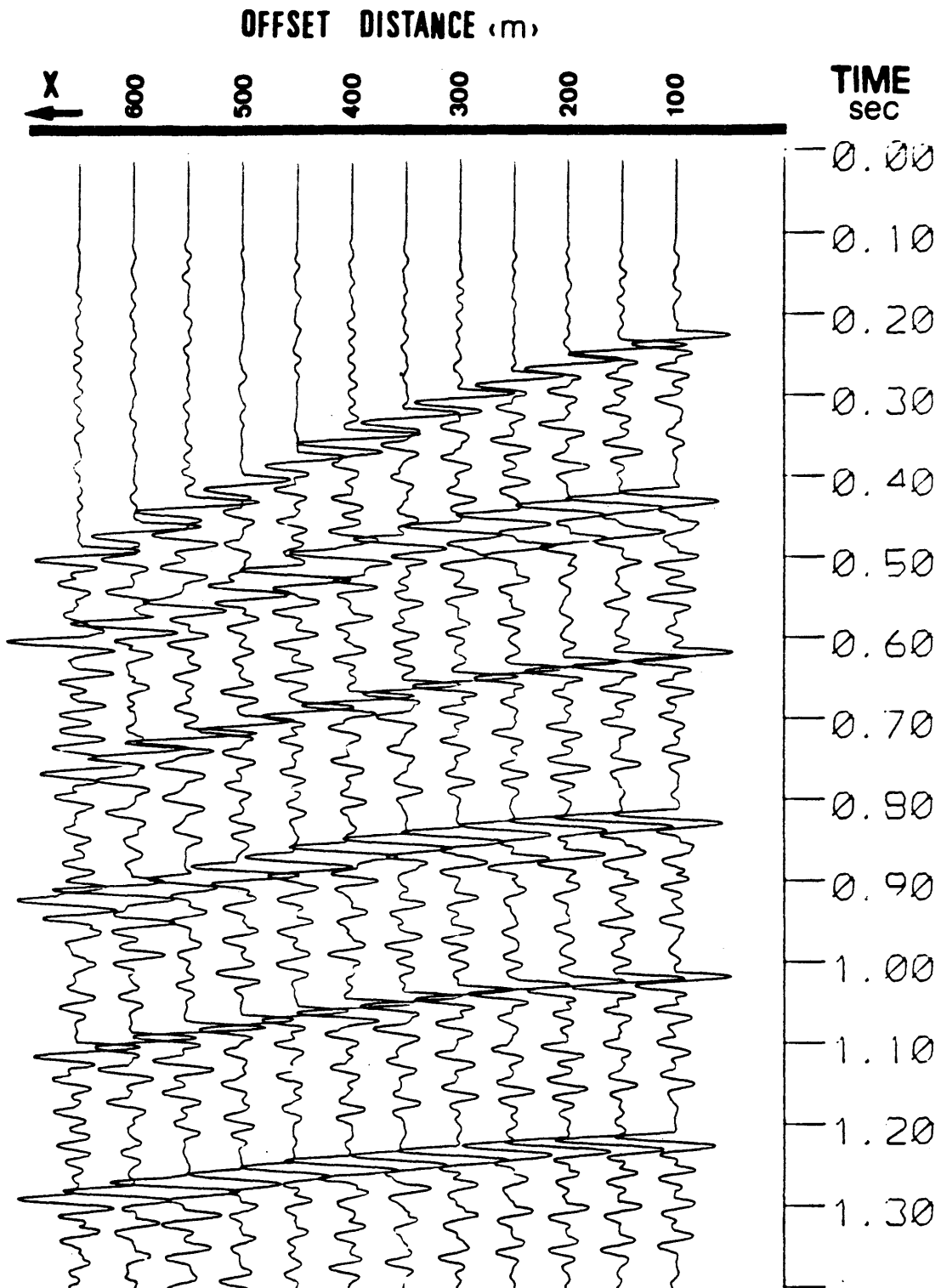


Figure 4-25, Prestack deghosting result.

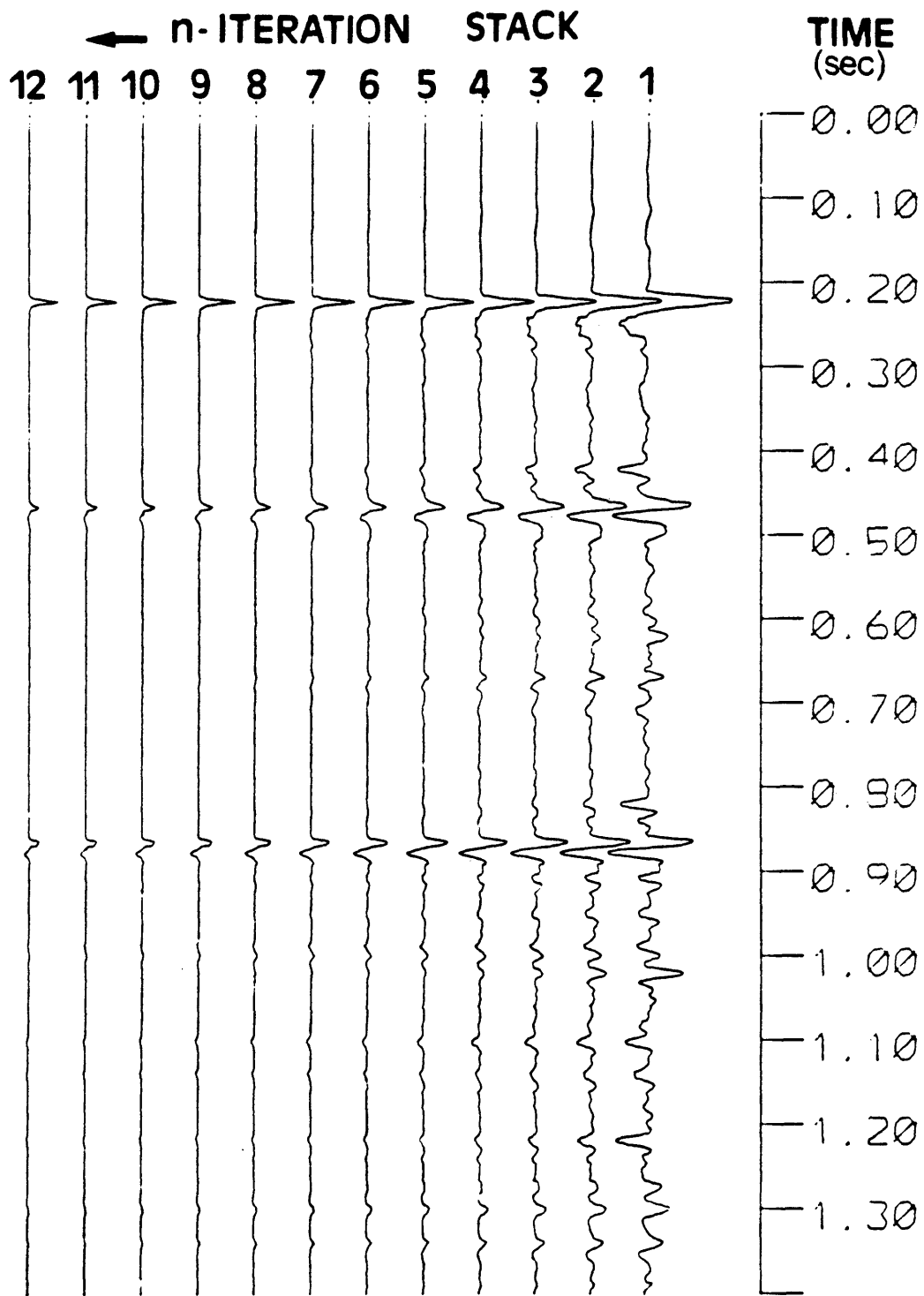


Figure 4-26, The final section after iterative process on figure 4-25.

5. CONCLUSIONS

From all of our modelling tests and comparisons, we have successfully fulfilled the objectives of understanding the characteristics of the iterative stacking algorithm, and of finding an accurate way to conduct the iterative stack in routine seismic processing. We summarized our conclusions as follows:

1) The iterative stack generally will be able to improve the primary to noise ratio because the algorithm takes into account the horizontal consistency of the amplitudes and not only the total straight sum. The iterative stack has proven to be a powerful tool in cleaning up the random noise, and is especially well adapted to suppressing strong long-period waterbottom multiples.

2) The iterative algorithm is a very flexible method. By varying the number of iterations and the muting parameters, it is possible to get a varying degree of emphasis on the horizontal consistency of the input. The algorithm may be considered as a generalized stacking method. In fact the straight stack is the 1-iteration version of the iteration output.

With the algorithm it is possible to output several versions. Routinely one may choose for comparison purposes, to display the first stack in addition to the iterative stacked sections. The iterative computation is so fast that the extra cost of doing this will be only the display costs.

3) The iterative algorithm may be applied anywhere after the linear seismic processing where a straight summation of traces is performed. Examples are velocity analysis, vertical stack, and the CDP stack. In velocity analysis, the iterative algorithm may result in improved velocity determination, which again will improve the final processed data.

4) For high resolution purposes, single-trace-processing is suggested for the output of the iterative process. The signal-to-noise ratio will be improved and the high frequency content will not be degraded as much.

5) It is suggested that muting should be included with the process of the iterative stack. The objective of muting is to alleviate the influence from NMO stretching, head waves, and the reflection amplitude change versus incidence angle.

These factors affect the ordinary stack as well as the iterative stack.

6) From our modelling tests, the iterative stack is generally more effective in suppressing background noise than the optimum weighted stack; moreover, the optimum weighted stack cannot achieve its full effectiveness in practice. These two algorithms are only partially compatible. If we put the iterative stack in routine processing, it is better to keep the manual editing and discard the diversity stack of the optimum weighted process.

7) The comparison of the iterative stack and f-k filtering has shown that both methods work well in deleting the horizontally travelling noise (low velocity events); but f-k filtering seems not to be able to clean up multiples as well as the iterative stack does. An example shows that these two methods are not compatible if conducted consecutively. Aliasing and truncation problems which usually hamper the f-k transformation will make the combined output worse.

8) The iterative stack cannot suppress ghosts or water

reverberations. Deconvolution is still the best tool to suppress this noise. If the iterative stack is conducted prior the deconvolution, the Wiener operator could not be properly designed from stacked traces; muting and single-trace-processing may help reduce the error of the Wiener operator design for this approach. However, the much more accurate way is to do deconvolution prior to the iterative stack.

The examples show the disadvantages of the use of the iterative process. Iterative stacking changes the waveforms by its non-linear nature; the more iterations performed, the worse the distortion. An inference that can be drawn from this is that, when possible, linear processes should precede non-linear processes in the processing sequence.

9) Although some real data has been tested by Naess (1979), additional real data tests are still recommended. Other interesting topics are:

(i). The use of the iterative stack in the migration by summation method (Bortfeld 1974),

(ii). The susceptibility of the iterative stack to static errors.

In brief, iterative stacking is an exceptionally effective tool for the detection of coherent but weak signals; however it must be used with care. Most of our seismic data processing is based upon the approximate linearity of much of the physical process. If iterative stacking precedes such linearity-based processing it can invalidate the fundamental assumptions of the linearity-based processing.

6. REFERENCES

- Bortfeld, R., 1974, Methods and Trends in Modern seismic exploration: Prakla-Seismos Publication.
- Claerbout, J. F., 1976, Fundamentals of Geophysical Data Processing: with Applications to Petroleum Prospecting: New York, McGraw-Hill.
- Dongara, J. J., Moler, C. B., Bunch, J.R., and Stewart G. W., 1979, Linpack Users' Guide: Society for Industrial and Applied Mathematics, Philadelphia.
- George, C. F., Jr., Hall, S. D., and Slack, H. A., 1968, Coherence Measurements and Their Application in the Analysis of Seismic Data: Presented at the 38th Annual International SEG Meeting, Denver, Colorado.
- GRAY Suite, 1983, The Generalized Ray Programs, Edited by Dr. F. Hadsell, Colorado School of Mines.
- Green, P. E., Kelly, E. J., and Levin, M. J., 1966, A Comparison of Seismic Array Processing Methods: Geophys. J. Ray. Astr. Soc., 11, 64-84.
- Lee, Y. W., 1960, Statistical Theory of Communication: New York, John Wiley and Sons, Inc.
- Mayne, W. H., 1962, Common Reflection Point Horizontal Stacking Technique: Geophysics 27, 92-114.
- Naess, O. E., 1979, Superstack- an iterative stacking algorithm: Geophysical Prospecting 27, 16-28.
- Naess, O. E., and Bruland, L., 1981, Velocity analysis using iterative stacking: Geophysical Prospecting 29, 1-20.
- Naess, O. E., 1982, Single-trace processing using iterative CDP-stacking: Geophysical Prospecting 30, 641-652.
- Neidell, N. S., and Taner, M. T., 1971, Semblance and other coherency measure for multichannel data: Geophysics 36, 482-497.
- Ngian, F. H., 1983, Personal Communication.
- Peacock, K. L., and Treitel, S., 1969, Predictive

Deconvolution, Theory and Practice: Geophysics 34, 155-169.

Rice, R. B., 1962, Inverse Convolution Filters: Geophysics 27, 4-18.

Robinson, E. A., 1966, Multichannel Z Transforms and Minimum Delay: Geophysics 31, 482-500.

Robinson, E. A., and Treitel, S., 1967, Principles of Digital Wiener Filtering, : Geophysical Prospecting 15, 311-333.

Robinson, J. C., 1970, Statistically optimal stacking of seismic data: Geophysics 35, 436-446.

SEG 1974, Digital Processing of Geophysical Data-a Review, Continuing Education Program: Society of Exploration Geophysicists.

Sengbush, R. L., 1983, Seismic Exploration Methods: IHRDC Publishers, 137 Newbury Street, Boston, MA 02116.

Sheriff, R. E., 1982, Encyclopedia Dictionary of Exploration Geophysics. SEG, Tulsa, Oklahoma.

White, R. E., 1977, The performance of optimum stacking filters in suppressing uncorrelated noise: Geophysical Prospecting 25, 165-178.

7. APPENDICES

APPENDIX A

MAIN PROGRAMS

Main programs to produce a printout, an output file, and/or a plot. (time unit: second, length unit: meter)

- The Program 'STKTST'
- The Program 'YMOD'
- The Program 'XMOD'
- The Program 'FTMAIN'
- The Program 'DECON'
- The Program 'PDECON'
- The Program 'TRPLOT'
- The Program 'PLOFIL'


```

00010          PROGRAM STKTST
00020 C*****
00030 C
00040 C          STKTST TESTS THE PERFORMANCE OF THE
00050 C          ITERATIVE ALGORITHM.
00060 C
00070 C          ARRAY:
00080 C          X=10 RANDOM DATA.
00090 C
00100 C          YEN MAR-1984
00110 C*****
00120 C
00130          DIMENSION X(10),BUF(10)
00140          TYPE 150
00150          150  FORMAT(1X,'          10 RANDOM DATA TEST',/)
00160 C*****SET RANDOM DATA
00170          DO 1 I=1,10
00180          X(I)=RAN(N)
00190          1  CONTINUE
00200 C*****SET ITERATION TIMES
00210          NITER=20
00220 C*****MAKE ITERATIVE STACK AND PRINTOUT
00230          DO 60 ITER=1,NITER
00240          TPOS=0.0
00250          TNEG=0.0
00260          DO 70 J=1,N
00270          IF(X(J).GT. 0.0) TPOS=TPOS+X(J)
00280          70  IF(X(J).LT. 0.0) TNEG=TNEG+X(J)
00290          XN=N
00300          TPOS=TPOS/XN
00310          TNEG=TNEG/XN
00320          TT=TPOS+TNEG
00330          TYPE 100,X,TPOS
00340          100  FORMAT(1X,10F7.2,7X,'TPOS=',F5.2)
00350          TYPE 200,TNEG
00360          200  FORMAT(1X,77X,'TNEG=',F5.2)
00370          TYPE 300,ITER,TT
00380          300  FORMAT(1X,57X,I2,'-ITERATION OUTPUT--',F5.2/)
00390          DO 75 J=1,N
00400          IF(X(J).GT.TPOS) X(J)=TPOS
00410          75  IF(X(J) .LT.TNEG) X(J)=TNEG
00420          60  CONTINUE
00430          STOP
00440          END
*
```

```

00010          PROGRAM YMOD
00020 C*****
00030 C
00040 C          YMOD GENERATES 12-FOLD CDP-GATHER FROM THE
00050 C          3-LAYER MODEL OR THE 4-LAYER MODEL.
00060 C
00070 C          ARRAYS:
00080 C          TC= 12-FOLD CDP-GATHER TRACES.
00090 C          X =OFFSET DISTANCE.
00100 C          XV=(X/VRMS)**2
00110 C          SW=SOURCE WAVELET.
00120 C          V =VRMS VELOCITY.
00130 C
00140 C          YEN MAR-1984
00150 C*****
00160 C          DIMENSION TC(700,12),X(12),XV(12),SW(40),V(700)
00170 C*****SET SINWAVE
00180 C          CALL SINWAV(41.,0.002,SW,25)
00190 C*****SET 3-LAYER MODEL
00200 C          CALL MODA(TC,X,XV)
00210 C*****YOU CAN ALSO CALL MODB FOR 4-LAYER MODEL.
00220 C*****ADD RANDOM NOISE
00230 C          CALL RAM(TC,0.06)
00240 C*****ADD GROUND NOISE
00250 C          CALL GROUND(TC,X)
00260 C*****CONVOLVE TC WITH THE WAVELET
00270 C          CALL CONVTC(TC,SW,700,12,25)
00280 C*****NMO CORRECTION
00290 C          CALL NMO(TC,V,X)
00300 C*****YOU CAN PLOT TC TO GET MODEL TRACES.
00310 C*****YOU CAN ALSO STORE TC INTO A FILE.
00320 C*****MAKE ITERATIVE STACK
00330 C          CALL STK(TC,12,2)
00340 C*****PLOT THE RESULT STACKS
00350 C          CALL PLO(TC,0.33,12,700,700)
00360 C          STOP
00370 C          END
*
```

```

00010          PROGRAM XMOD
00020 C*****
00030 C
00040 C          XMOD COMPUTES A MODEL OF 6-FOLD CDP-GATHER
00050 C          TRACES AND CALCULATES ITS OPTIMUM WEIGHTED STACK
00060 C          AND ITERATIVE STACK.
00070 C
00080 C          ARRAYS:
00090 C          TC=TRACE 1-6:  6-FOLD CDP-GATHER.
00100 C                   7:  IDEAL SIGNAL TRACE.
00110 C                   8:  STRAIGHT STACK.
00120 C                   9:  THE 4-ITERATION STACK.
00130 C                   10: OPTIMUM WEIGHTED STACK.
00140 C                   11: THE 4-ITERATION AFTER WEIGHT.
00150 C
00160 C          YEN MAR-1984
00170 C*****
00180 C          DIMENSION TC(280,11),BUF(280,7),SW(40)
00190 C          DIMENSION RATE(7),SCL(7)
00200 C*****SET SIGNAL SCALE
00210 C          DATA SCL/1.,1.,1.,1.,1.,1.,1./
00220 C*****SET RICKER WAVELET
00230 C          CALL RICKER(41.,0.002,SW,40)
00240 C*****SET MODELLING TRACE 1-6
00250 C          CALL MODX(TC,BUF,SW,RATE,SCL)
00260 C*****MAKE ITERATIVE STACK
00270 C          CALL STKGEN(BUF,280,6,6,2)
00280 C*****TRANSFER THE STRAIGHT STACK AND THE
00290 C          4-ITERATION STACK INTO TC TRACE 8 & 9
00300 C          DO 1 I=1,280
00310 C             TC(I,8)=BUF(I,1)
00320 C             1 TC(I,9)=BUF(I,4)
00330 C*****MAKE OPTIMUM WEIGHTING
00340 C          CALL WTD(TC,BUF,RATE,SCL)
00350 C*****MAKE ITERATIVE STACK ON WEIGHTED TRACES
00360 C          CALL STKGEN(BUF,280,6,6,2)
00370 C*****TRANSFER THE OPTIMUM WEIGHTED STACK AND
00380 C          THE 4-ITERATION WEIGHTED STACK INTO TC 10 & 11
00390 C          DO 2 I=1,280
00400 C             TC(I,10)=BUF(I,1)
00410 C             2 TC(I,11)=BUF(I,4)
00420 C*****PLOT THE 11 TRACES
00430 C          CALL PLOTX(TC,RATE,SCL)
00440 C          STOP
00450 C          END
*
```

```
00010          PROGRAM FTMAIN
00020 C*****
00030 C
00040 C          FTMAIN DOES F-K FILTERING AND PLOTS THE RESULT.
00050 C          ITS INPUT IS READ FROM FOR02.DAT.
00060 C
00070 C          VARIABLES:
00080 C          DIP=DIPPING ANGLE IN DEGREES FOR REJECT ZONE.
00090 C          ARRAYS:
00100 C          TC =12-FOLD CDP-GATHER.
00110 C
00120 C          YEN MAR-1984
00130 C*****
00140          DIMENSION TC(700,12)
00150 C*****SET REJECT DIP
00160          DIP=45.0
00170 C*****READ 12-FOLD TRACES FROM FILE FOR02.DAT
00180          DO 1 J=1,12
00190             1 READ(2,100) (TC(I,J),I=1,700)
00200 C*****EXEC F-K FILTERING
00210          CALL FT(TC,DIP,12)
00220             100 FORMAT(10E11.4)
00230 C*****PLOT THE RESULT
00240          CALL PLO(TC,0.33,12,512,700)
00250          STOP
00260          END
*
```

```

00010          PROGRAM DECON
00020 C*****
00030 C
00040 C          DECON APPLIES THE WIENER PROCESS TO INPUT ARRAY X.
00050 C          OUTPUT IS THE PLOT OF 12-FOLD DECON TRACES.
00060 C
00070 C          VARIABLES:
00080 C          LDA=LEADING DIMENSION OF MATRIX A.
00090 C          N=FILTER LENGTH.
00100 C          NZ=0, IF TRACES ARE NMO-CORRECTED.
00110 C          OTHERWISE NOT YET.
00120 C          ARRAYS:
00130 C          X = INPUT ARRAY IN FILE FOR02.DAT WHICH
00140 C          CONTAINS 12-FOLD CDP-GATHER WITH 700
00150 C          DATA IN EACH TRACE; X WILL BE DECONVOLVED
00160 C          AFTER PROCESSED.
00170 C          A = MATRIX TO COMPUTE THE FILTER.
00180 C          Z = DESIRED OUTPUT TRACE.
00190 C          XX= AUTOCORRELATION OF X.
00200 C          ZX= CROSSCORRELATION OF Z AND X.
00210 C          IPVT=DUMMY FOR SUBROUTINE SGEFA & SGESL.
00220 C          TC= 12-FOLD CDP-GATHER.
00230 C
00240 C          YEN MAR-1984
00250 C*****
00260 C
00270 C          DIMENSION TC(700,12),A(120,120),SW(40)
00280 C          DIMENSION X(700),Z(700),XX(700),ZX(700),IPVT(120)
00290 C*****SET PARAMETERS
00300 C          LDA=120
00310 C          N=45
00320 C          NZ=0
00330 C*****SET SINWAVE
00340 C          CALL SINWAV(40.,0.002,SW,20)
00350 C          DO 1 L=1,12
00360 C*****READ IN ONE TRACE
00370 C          READ(2,100) (X(K)=1,700)
00380 C*****SET TRACE WINDOW
00390 C          CALL TRSET(X,L,NST,NW,NZ)
00400 C*****SET DESIRED OUTPUT
00410 C          Z(4)=1.0
00420 C*****CHECK THE WINDOW PARAMETERS
00430 C          TYPE 200,L,NST,NW
00440 C*****COMPUTE AUTOCORRELATION OF X
00450 C          CALL CROSS(X,X,XX,NW,NW)
00460 C*****COMPUTE CROSS CORRELATION OF ZX
00470 C          CALL CROSS(Z,X,ZX,NW,NW)
00480 C*****SET MATRIX
00490 C          CALL MXSET(XX,N,A,LDA)
00500 C*****SOLVE MATRIX
00510 C          CALL SGEFA(A,LDA,N,IPVT,INFO)
00520 C          IF(INFO.NE.0) GOTO 99
00530 C          CALL SGESL(A,LDA,N,IPVT,ZX,0)
00540 C*****SPIKE DECON
00550 C          CALL CONVXF(X,ZX,X,700,N,700)
00560 C          DO 3 I=1,700
00570 C          3   TC(I,L)=X(I)
00580 C          1   CONTINUE
00590 C          GOTO 999

```

```
00600      99      TYPE 110
00610      110     FORMAT(1X,'INFO NOT 0')
00620      999     CONTINUE
00630      C*****CONVOLVE TC WITH KNOWN WAVELET
00640      CALL CONVTC(TC,SW,700,12,20)
00650      CALL PLO(TC,0.25,12,700,700)
00660      100     FORMAT(10E11.4)
00670      200     FORMAT(1X,'L=',I2,'NST=',I3,'NW=',I3)
00680      STOP
00690      END
*
```

```

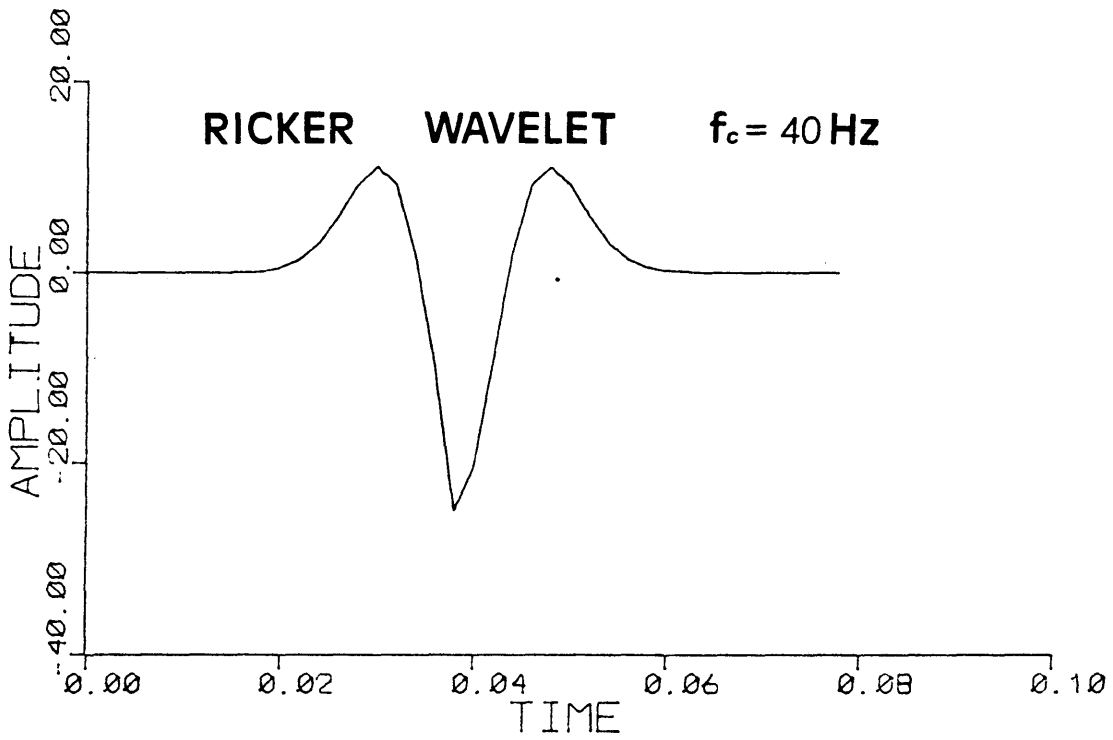
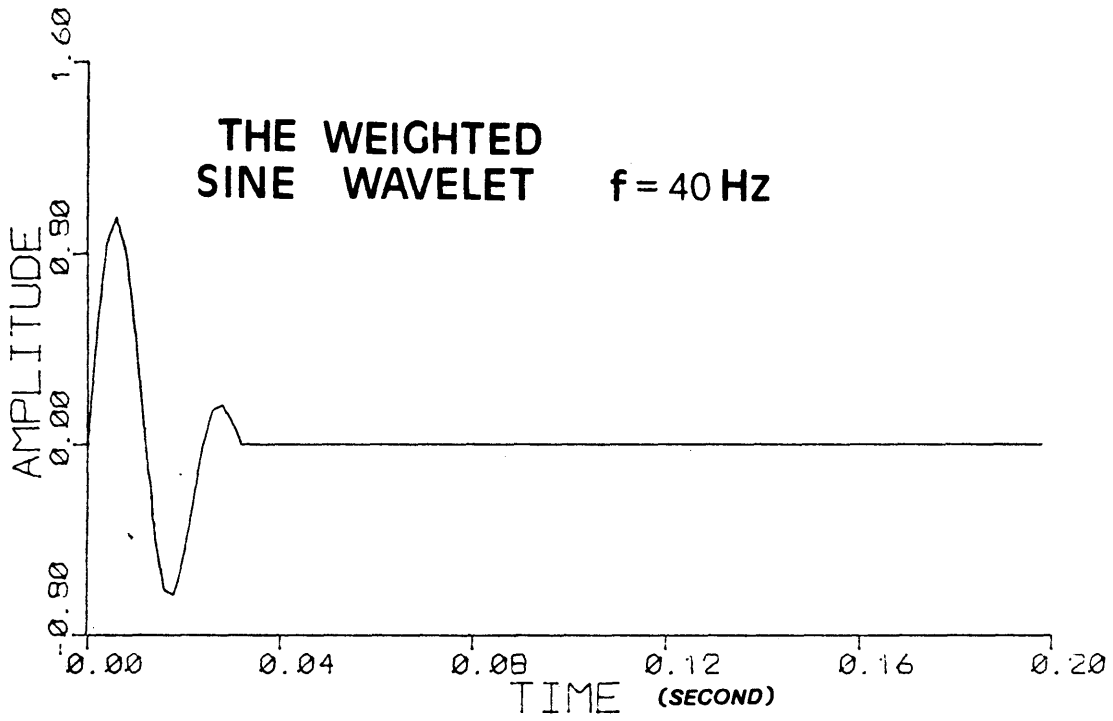
00010      PROGRAM PDECON
00020      C*****
00030      C
00040      C      PDECON DOES PREDICTIVE DECON ON INPUT ARRAY X.
00050      C      INPUT ARRAY READ FROM FOR02.DAT AND,
00060      C      OUTPUT IS THE PLOT OF 12-FOLD DECON TRACES.
00070      C
00080      C      VARIABLES:
00090      C      LDA=LEADING DIMENSION OF MATRIX A.
00100      C      N =LENGTH OF THE PREDICTIVE FILTER.
00110      C      NZ=0, IF TRACES ARE NMO-CORRECTED.
00120      C      OTHERWISE NOT YET.
00130      C      NR=PREDICTIVE DISTANCE.
00140      C      NF=LENGTH OF THE PREDICTIVE ERROR FILTER.
00150      C      ARRAYS:
00160      C      X = INPUT ARRAY IN FILE FOR02.DAT WHICH
00170      C      CONTAINS 12-FOLD CDP-GATHER WITH 700
00180      C      DATA IN EACH TRACE; X WILL BE DECONVOLVED
00190      C      AFTER PROCESSED.
00200      C      A = MATRIX TO COMPUTE THE FILTER.
00210      C      Z = DESIRED OUTPUT TRACE.
00220      C      XX= AUTOCORRELATION OF X.
00230      C      ZX= CROSSCORRELATION OF Z AND X.
00240      C      IPVT=DUMMY FOR SUBROUTINE SGEFA & SGESL.
00250      C      TC= 12-FOLD CDP-GATHER.
00260      C
00270      C      YEN MAR-1984
00280      C*****
00290      C
00300      C      DIMENSION TC(700,12),A(120,120)
00310      C      DIMENSION X(700),Z(700),XX(700),ZX(700),IPVT(120)
00320      C*****SET PARAMETERS
00330      C      LDA=120
00340      C      N=45
00350      C      NZ=0
00360      C      NR=20
00370      C      NF=N+NR
00380      C*****START PREDIC DECON
00390      C      DO 1 L=1,12
00400      C*****INPUT TRACE
00410      C      READ(2,100) (X(K)=1,700)
00420      C*****SET TRACE WINDOW
00430      C      CALL TRSET(X,L,NST,NW,NZ)
00440      C*****CHECK WINDOW PARAMETERS
00450      C      TYPE 200,L,NST,NW
00460      C*****COMPUTE AUTO X
00470      C      CALL CROSS(X,X,XX,NW,NW)
00480      C*****SET PREDICTIVE OUTPUT
00490      C      CALL PRDZX(XX,ZX,NW,NR)
00500      C*****SET MATRIX
00510      C      CALL MXSET(XX,N,A,LDA)
00520      C*****SOLVE MATRIX
00530      C      CALL SGEFA(A,LDA,N,IPVT,INFO)
00540      C      IF(INFO.NE.0) GOTO 99
00550      C      CALL SGESL(A,LDA,N,IPVT,ZX,0)
00560      C*****SET PRED ERROR FILTER
00570      C      CALL FSET(ZX,NR,N)
00580      C*****PREDIC DECON
00590      C      CALL CONVXF(X,ZX,X,700,NF,700)

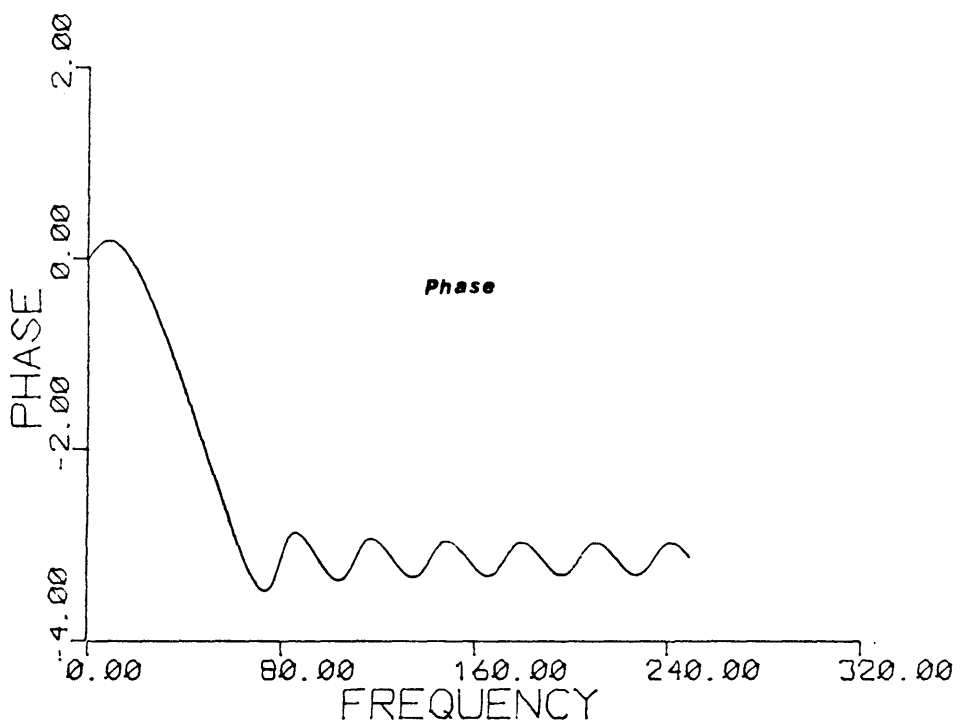
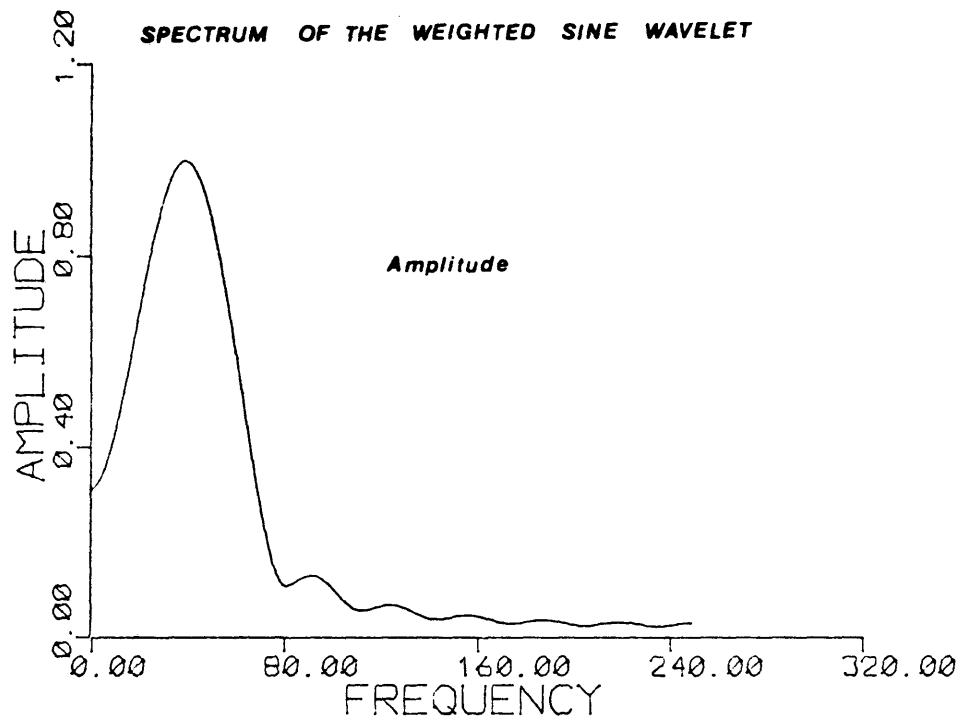
```

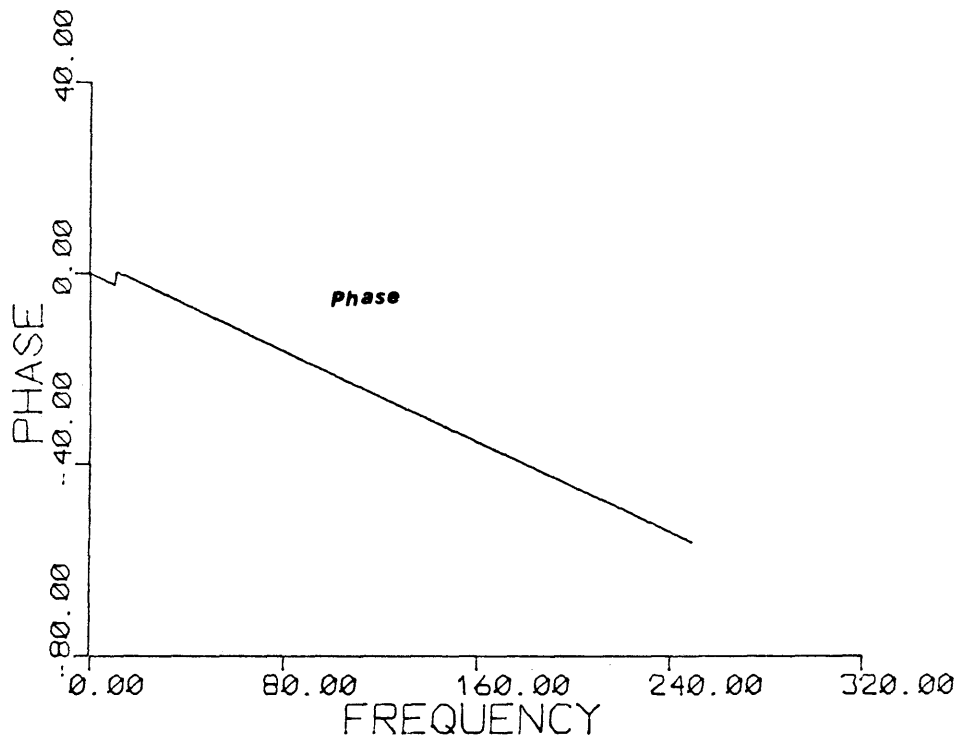
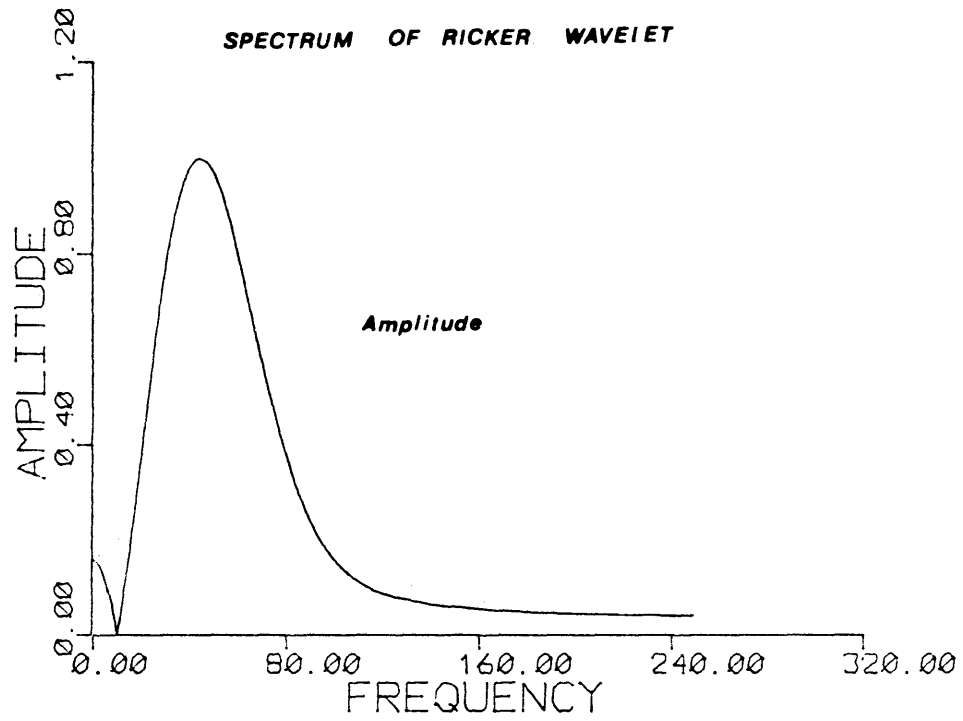
```
00600 C*****TRANSFER DECON TRACE TO TC
00610          DO 3 I=1,700
00620          3   TC(I,L)=X(I)
00630          1   CONTINUE
00640          GOTO 999
00650          99   TYPE 110
00660          110  FORMAT(1X,'INFO NOT 0')
00670          999  CONTINUE
00680 C*****PLOT TC
00690          CALL PLO(TC,0.25,12,700,700)
00700          100  FORMAT(10E11.4)
00710          200  FORMAT(1X,'L=',I2,'NST=',I3,'NW=',I3)
00720          STOP
00730          END
*
```



```
00010          PROGRAM TRPLOT
00020 C*****
00030 C
00040 C      TRPLOT PLOTS A TRACE IN TIME DOMAIN OR TRANSFORMS
00050 C      THE TRACE INTO FREQUENCY DOMAIN AND PLOTS
00060 C      ITS AMPLITUDE AND PHASE SPECTRUMS.
00070 C
00080 C      ARRAY:
00090 C          FILT=THE TRACE TO BE PLOTTED.
00100 C
00110 C      YEN MAR-1984
00120 C*****
00130 C
00140 C          DIMENSION FILT(700)
00150 C*****SET SINWAVE AND PLOT
00160 C          CALL SINWAV(40.0,0.002,FILT,16)
00170 C          CALL TRACE(100,FILT,0.002)
00180 C*****TRANSFORM AND PLOT ITS SPECTRUM
00190 C          CALL SPECPL(40,FILT,0.002)
00200 C*****SET RICKER WAVELET AND PLOT
00210 C          CALL RICKER(40.,0.002,FILT,40)
00220 C          CALL TRACE(100,FILT,0.002)
00230 C*****TRANSFORM AND PLOT ITS SPECTRUM
00240 C          CALL SPECPL(40,FILT,0.002)
00250 C          STOP
00260 C          END
*
```







```

00010          PROGRAM PLOFIL
00020 C*****
00030 C
00040 C          PLOFIL PLOTS A DATA FILE OF 12 TRACES
00050 C          IN INTERACTIVE MODE.
00060 C
00070 C          VARIABLES:
00080 C          NF=FILE NUMBER OF FOR00.DAT
00090 C          NPT=DATA POINT TO BE PLOTTED IN ONE TRACE.
00100 C          DX=SPACING BETWEEN TWO TRACES.
00110 C          ARRAYS:
00120 C          TC=ARRAY TO BE PLOTTED.
00130 C
00140 C          YEN MAR-1984
00150 C*****
00160          DIMENSION TC(700,12)
00170          TYPE 100
00180          100  FORMAT(1X,'INPUT FILE NUMBER')
00190          ACCEPT 150,NF
00200          150  FORMAT(I2)
00210          DO 1 J=1,12
00220          1    READ(NF,900) (TC(I,J), I=1,700)
00230          TYPE 200
00240          200  FORMAT(1X,'INPUT NPT')
00250          ACCEPT 250,NPT
00260          250  FORMAT(I3)
00270          TYPE 300
00280          300  FORMAT(1X,'INPUT DX')
00290          ACCEPT 350,DX
00300          350  FORMAT(F5.2)
00310          CALL PLO(TC,DX,12,NPT,700)
00320          900  FORMAT(10E11.4)
00330          STOP
00340          END
*
```

APPENDIX B

SUBROUTINES

Subroutines to support the execution of the main programs in modelling, processing, and plotting jobs.

THE LIST OF SUBROUTINES:

Modelling routines:

MODA(TC, X, XV)
MODE(TC, X, XV)
MODX(TC, BUF, SW, RATE, SCL)
RICKER(F, DT, FILT, NFF)
SINWAV(F, DT, SW, NFF)
RAM(TC, FACT)
GHOST(TC, ND, R)
GROUND(TC, X)
CONVTC(TC, SW, NTC, NL, NSW)

Processing routines:

CONVXF(A, FILT, C, NA, NFILT, NC)
CROSS(A, FILT, C, MA, MFILT)
CVS(TC, X, ITER, VSTR, VINC, VEND)
FFT(LX, CX, SIGNI)
FSET(F, NR, N)
FT(T, DIP, NL)
MXSET(XX, N, A, LDA)
NMD(TC, V, X)
PRDZX(XX, ZX, NX, NR)
SGEFA(A, LDA, N, IPVT, INFO)
SGESL(A, LDA, N, IPVT, B, JOB)
TRSET(X, L, NST, NW, NZ)
WTD(TC, BUF, RATE, SCL)

Iterative processing routines:

STK(TC, NITER, JOB)
STKGEN(TC, NPT, NN, NITER, JOB)
STKLEV(X, N, NITER, JOB)

Plotting routines:

FLO(T, DX, NL, NPT, LDM)
FLOLAB
PLOTX(T, RATE, SCL)
FLO2(T, DX, J)
SPECPL(NFF, FILT, DT)
TRACE(NFF, AMP, DT)

```

00010 C*****
00020 C
00030 C      SUBROUTINE CONVTC
00040 C
00050 C*****
00060 C
00070 C      CONVOLVES 2-D ARRAY TC WITH 1-D ARRAY SW.
00080 C
00090 C      VARIABLES:
00100 C          NTC=DATA POINTS OF EACH TRACE.
00110 C          NL =NUMBER OF TRACES TO BE CONVOLVED.
00120 C          NSW=DATA POINTS OF WAVELET.
00130 C
00140 C      ARRAYS:
00150 C          TC=12-FOLD CDP-GATHER TRACES, WHICH CONTAIN
00160 C          REFLECTIVITIES ON ENTRY, AND WILL
00170 C          BE SYNTHETIC TRACES ON RETURN.
00180 C          SW=SOURCE WAVELET.
00190 C          BUFF=BUFFER.
00200 C
00210 C      YEN MAR-1984
00220 C*****
00230
00240          SUBROUTINE CONVTC(TC,SW,NTC,NL,NSW)
00250          DIMENSION TC(NTC,1),SW(1),BUFF(700)
00260 C*****START CONVOLUTION
00270          DO 1 J=1,NL
00280          DO 2 I=2,NTC
00290 C*****SET WINDOW FOR MULTIPLICATION
00300          MT=NSW
00310          IA=I
00320          IB=1
00330          SUM=0.0
00340          IF(MT.GT.I) MT=I
00350 C*****MULTIPLY AND SUM INTO BUFFER
00360          DO 3 NN=1,MT
00370          SUM=SUM+TC(IA,J)*SW(IB)
00380          IA=IA-1
00390          3  IB=IB+1
00400          2  BUFF(I)=SUM
00410 C*****TRANSFER BUFFER TO TRACE TC.
00420          DO 4 I=1,NTC
00430          4  TC(I,J)=BUFF(I)
00440          1  CONTINUE
00450          TYPE 99
00460          99  FORMAT(1X,'CONV DONE')
00470          RETURN
00480          END
00490
*E

```



```

00010 C*****
00020 C
00030 C      SUBROUTINE CONVXF
00040 C
00050 C*****
00060 C
00070 C      CONVOLVES ARRAY A AND FILTER B.
00080 C
00090 C      VARIABLES:
00100 C          NA=LENGTH OF ARRAY A.
00110 C          NB=LENGTH OF FILTER B.
00120 C          NC=LENGTH OF RESULTED TRACE.
00130 C      ARRAYS:
00140 C          A=ARRAY A ON ENTRY.
00150 C          B=FILTER B ON ENTRY.
00160 C          C=TO BE RETURN.
00170 C          BUFF=BUFFER TO AVOID DAMAGE WHEN C IS A.
00180 C
00190 C      YEN MAR-1984
00200 C*****
00210 C      SUBROUTINE CONVXF(A,B,C,NA,NB,NC)
00220 C          DIMENSION A(1),B(1),C(1),BUFF(700)
00230 C*****SET LENGTH OF NC
00240 C          NC=NA+NB-1
00250 C          IF(NC.GT.700) NC=700
00260 C*****START CONVOLUTION
00270 C          DO 1 I=1,NC
00280 C*****SET CONVOLUTION WINDOW
00290 C          MT=NB
00300 C          IA=I
00310 C          IB=1
00320 C          SUM=0.0
00330 C          IF(MT .GT. I) MT=I
00340 C*****MULTIPLY AND SUM
00350 C          DO 2 NN=1,MT
00360 C              SUM=SUM+A(IA)*B(IB)
00370 C              IA=IA-1
00380 C              IB=IB+1
00390 C              1  BUFF(I)=SUM
00400 C*****TRANSFER FROM BUFFER TO C
00410 C          DO 3 I=1,NC
00420 C              3  C(I)=BUFF(I)
00430 C          RETURN
00440 C          END
*EU

```

```

00010 C*****
00020 C
00030 C      SUBROUTINE CROSS
00040 C
00050 C*****
00060 C
00070 C      COMPUTES CROSS CORRELATION OF A AND B.
00080 C
00090 C      VARIABLES:
00100 C          MA=NUMBER OF ELEMENTS IN A.
00110 C          MB=NUMBER OF ELEMENTS IN B.
00120 C      ARRAYS:
00130 C          A = ENTRY TRACE, A IS LONGER THAN B.
00140 C          B = ENTRY TRACE TO BE CROSS WITH A.
00150 C          C = RETURN TRACE WITH LENGTH OF MA.
00160 C          BUF=BUFFER IN CASE C IS A.
00170 C
00180 C      YEN MAR-1984
00190 C*****
00200 C      SUBROUTINE CROSS(A,B,C,MA,MB)
00210 C          DIMENSION A(1),B(1),C(1),BUF(700)
00220 C          IF(MA.GT.700) MA=700
00230 C*****START CONVOLUTION
00240 C          DO 1 I=1,MA
00250 C              IA=I
00260 C              IB=1
00270 C              SUM=0.0
00280 C*****MULTIPLY AND SUM
00290 C          DO 2 K=I,MB
00300 C              IF(IS.GT.700) GOTO 2
00310 C              SUM=SUM+A(IA)*B(IB)
00320 C              IA=IA+1
00330 C          2      IB=IB+1
00340 C              BUF(I)=SUM
00350 C          1      CONTINUE
00360 C*****TRANSFER BUFFER TO C
00370 C          DO 3 I=1,MA
00380 C          3      C(I)=BUF(I)
00390 C          RETURN
00400 C          END
*EU

```

```

00010 C*****
00020 C
00030 C      SUBROUTINE CVS
00040 C
00050 C*****
00060 C
00070 C      PROCESSES CONSTANT VELOCITY ANALYSES
00080 C      BY ITERATIVE STACKING AND PLOT.
00090 C
00100 C      VARIABLES:
00110 C          ITER=NUMBER OF ITERATIONS WHEN MAKE STACKING,
00120 C          THE OUTPUT WILL BE STRAIGHT STACK IF 1.
00130 C          VSTR=BEGINNING VELOCITY ON TEST.
00140 C          VINT=INCREMENT IN VELOCITY TEST.
00150 C          VEND=ENDING VELOCITY ON TEST.
00160 C      ARRAYS:
00170 C          TC=12-FOLD CDP-GATHER.
00180 C          X =OFFSET**2
00190 C          XNMD=ONE LEVEL NMD-CORRECTED DATA.
00200 C          TV=CVS TRACE TO BE PLOTTED.
00210 C
00220 C      YEN MAR-1984
00230 C*****
00240 C      SUBROUTINE CVS(TC,X,ITER,VSTR,VINC,VEND)
00250 C      DIMENSION TC(700,1),X(1),XNMD(12),TV(700)
00260 C*****COMPUTE HOW MANY CVS' TO BE TESTED
00270 C      NL=((VEND-VSTR)/VINC)+1
00280 C      DO 1 J=1,NL
00290 C*****SET CONSTANT VELOCITY
00300 C      VAG=(VSTR+(J-1)*VINC)**2
00310 C      DO 2 I=1,700
00320 C*****SET ZERO OFFSET TIME
00330 C      TZ=(0.002*I)**2
00340 C      DO 3 L=1,12
00350 C*****FIND OFFSET ARRIVING TIME
00360 C      AND MAKE NMD CORRECTION
00370 C      XV=X(L)/VAG
00380 C      NTX=SQRT(TZ+XV)*500.0+1
00390 C      IF(NTX.GT.700) NTX=700
00400 C      3  XNMD(L)=TC(NTX,L)
00410 C*****MAKE ITERATIVE STACK
00420 C      CALL STKLEV(XNMD,12,ITER,2)
00430 C*****TRANSFER ITER-STACK INTO TV
00440 C      2  TV(I)=XNMD(ITER)
00450 C*****PLOT TV
00460 C      1  CALL PLO2(TV,DX,J)
00470 C      CALL PLOT(0,0,+999)
00480 C      RETURN
00490 C      END
*EU

```

```

00010 C*****
00020 C
00030 C      SUBROUTINE FFT
00040 C
00050 C*****
00060 C
00070 C      COMPUTES A FAST FOURIER TRANSFORMATION.
00080 C
00090 C      VARIABLES:
00100 C          LX= NUMBER OF DATA POINTS IN ARRAY.
00110 C          SIGNI= +1. TO GO FROM TIME TO FREQUENCY,
00120 C                -1 TO GO FROM FREQUENCY TO TIME.
00130 C      ARRAYS:
00140 C          CX= ARRAY TO BE TRANSFORMED.
00150 C
00160 C      CLEARBOUT 1976, AFTER BRENNER
00170 C*****
00180 C      SUBROUTINE FFT(LX,CX,SIGNI)
00190 C      COMPLEX CX(LX),CARG,CEXP,CW,CTEMP
00200 C      J=1
00210 C      SC=SQRT(1./LX)
00220 C      DO 30 I=1,LX
00230 C      IF(I.GT.J)GOTO 10
00240 C      CTEMP=CX(J)*SC
00250 C      CX(J)=CX(I)*SC
00260 C      CX(I)=CTEMP
00270 C      10  M=LX/2.
00280 C      20  IF(J.LE.M) GOTO 30
00290 C          J=J-M
00300 C          M=M/2
00310 C      IF(M.GE.1) GOTO 20
00320 C      30  J=J+M
00330 C          L=1
00340 C      40  ISTEP=2*L
00350 C          DO 50 M=1,L
00360 C          CARG=(0.,1.)*(3.14159265*SIGNI*(M-1))/L
00370 C          CW=CEXP(CARG)
00380 C          DO 50 I=M,LX,ISTEP
00390 C          CTEMP=CW*CX(I+L)
00400 C          CX(I+L)=CX(I)-CTEMP
00410 C      50  CX(I)=CX(I)+CTEMP
00420 C          L=ISTEP
00430 C          IF(L.LT.LX) GOTO 40
00440 C      RETURN
00450 C      END
*EU

```

```

00010 C*****
00020 C
00030 C      SUBROUTINE FSET
00040 C
00050 C*****
00060 C
00070 C      SETS PREDICTIVE ERROR FILTER.
00080 C
00090 C      VARIABLES:
00100 C          NR=PREDICTIVE LENGTH IN NUMBER OF DATA POINTS.
00110 C          N =LENGTH OF THE PREDICTIVE FILTER.
00120 C      ARRAYS:
00130 C          F =PREDICTIVE FILTER ON ENTRY.
00140 C          PREDICTIVE ERROR FILTER ON RETURN.
00150 C
00160 C      YEN MAR-1984
00170 C*****
00180 C      SUBROUTINE FSET(F,NR,N)
00190 C      DIMENSION F(1)
00200 C*****SHIFT VALUE DOWNAND TIMES -1.
00210 C      DO 1 I=1,N
00220 C          J=N+1-I
00230 C          JJ=J+NR
00240 C          F(JJ)=F(J)*(-1.0)
00250 C      1  F(J)=0.0
00260 C          IF(NR.EQ. 1) GOTO 3
00270 C          KK=NR-1
00280 C*****PUT ZERO IN THE GAP.
00290 C      DO 2 I=1,KK
00300 C          II=I+1
00310 C      2  F(II)=0.0
00320 C      3  F(1)=1.0
00330 C      RETURN
00340 C      END
*EU

```

```

00010 C*****
00020 C
00030 C      SUBROUTINE FT
00040 C
00050 C*****
00060 C
00070 C      MAKE F-K FILTERING ACCORDING TO
00080 C      THE DEGREE OF DIP ON ENTRY.
00090 C
00100 C      VARIABLES:
00110 C          DIP=THE SLOPE OF SLICING IN DEGREES.
00120 C          NL=NUMBER OF FOLDS FOR CDP-GATHER.
00130 C      ARRAYS:
00140 C          T =12-FOLD CDP-GATHER.
00150 C          AA=TRANSFORMATION FORM OF THE CDP-GATHER.
00160 C          A =VERTICAL TRANSFORMABLE TRACE FOR FFT.
00170 C          B =HORIZONTAL TRANSFORMABLE TRACE FOR FFT.
00180 C
00190 C      YEN MAR-1984
00200 C*****
00210 C      SUBROUTINE FT(T,DIP,NL)
00220 C          DIMENSION T(700,1)
00230 C          COMPLEX AA(512,16),A(512),B(16)
00240 C*****CLEAR AA.
00250 C          DO 1 J=1,16
00260 C              DO 2 I=1,512
00270 C                  2 AA(I,J)=CMPLX(0.0,0.0)
00280 C                  1 CONTINUE
00290 C*****SET X-T FORM FOR TRANSFORMATION.
00300 C          DO 3 J=1,NL
00310 C              DO 4 I=1,512
00320 C                  TT=T(I,J)
00330 C                  4 AA(I,J)=CMPLX(TT,0.0)
00340 C                  3 CONTINUE
00350 C*****F TRANSFORM.
00360 C          DO 5 J=1,16
00370 C              DO 6 I=1,512
00380 C                  6 A(I)=AA(I,J)
00390 C                  CALL FFT(512,A,1.)
00400 C                  DO 7 I=1,512
00410 C                      7 AA(I,J)=A(I)
00420 C                  5 CONTINUE
00430 C*****K TRANSFORM.
00440 C          DO 8 I=1,512
00450 C              DO 9 J=1,16
00460 C                  9 B(J)=AA(I,J)
00470 C                  CALL FFT(16,B,1.)
00480 C                  DO 10 J=1,16
00490 C                      10 AA(I,J)=B(J)
00500 C                  8 CONTINUE

```

```

00510 C*****ZERO-OUT IN F-K DOMAIN.
00520     DIP=DIP*3.141598/180.
00530     DIP=TAN(DIP)*256.0/9.0
00540     DO 11 J=2,16
00550     JJ=J
00560     IF(J.GT.9) JJ=18-J
00570     NN=JJ*DIP
00580     IF(NN.GT.256)NN=256
00590     DO 12 I=1,NN
00600     II=513-I
00610     AA(II,J)=CMPLX(0.0,0.0)
00620     12 AA(I,J)=CMPLX(0.0,0.0)
00630     11 CONTINUE
00640 C*****X TRANSFORM.
00650     DO 13 I=1,512
00660     DO 14 J=1,16
00670     14 B(J)=AA(I,J)
00680     CALL FFT(16,B,-1.0)
00690     DO 15 J=1,16
00700     15 AA(I,J)=B(J)
00710     13 CONTINUE
00720 C*****T TRANSFORM.
00730     DO 18 J=1,16
00740     DO 19 I=1,512
00750     19 A(I)=AA(I,J)
00760     CALL FFT(512,A,-1.0)
00770     DO 20 I=1,512
00780     20 AA(I,J)=A(I)
00790     18 CONTINUE
00800 C*****BACK TO ORIGINAL CDP-GATHER.
00810     DO 21 J=1,NL
00820     DO 22 I=1,512
00830     22 T(I,J)=REAL(AA(I,J))
00840     21 CONTINUE
00850     TYPE 99
00860     99 FORMAT(1X,'FT DONE')
00870     RETURN
00880     END
*
```

```

00010 C*****
00020 C
00030 C      SUBROUTINE GHOST
00040 C
00050 C*****
00060 C
00070 C      SETS GHOST RESPONSES.
00080 C
00090 C      VARIABLES:
00100 C          ND=NUMBER OF DATA POINTS FOR GHOST DELAY.
00110 C          R =REFLECTION COEFF OF THE GHOST.
00120 C      ARRAY:
00130 C          TC=12-FOLD CDP-GATHER.
00140 C
00150 C      YEN MAR-1984
00160 C*****
00170 C      SUBROUTINE GHOST(TC,ND,R)
00180 C          DIMENSION TC(700,1)
00190 C          DO 1 J=1,12
00200 C          DO 2 N=1,700
00210 C          I=701-N
00220 C          IGHOST=I+ND
00230 C          IF(IGHOST.GT.700) GOTO 2
00240 C          TC(IGHOST,J)=TC(IGHOST,J)+TC(I,J)
00250 C          2 CONTINUE
00260 C          1 CONTINUE
00270 C          RETURN
00280 C          END
*EU

```



```

00010 C*****
00020 C
00030 C      SUBROUTINE GROUND
00040 C
00050 C*****
00060 C
00070 C      GENERATES GROUND TRAVELLING NOISE.
00080 C
00090 C      VARIABLES:
00100 C          K=GROUND EVENTS FROM 1 TO 4 WITH VELOCITIES
00110 C          400,450,500,550 METER/SEC RESPECTIVELY.
00120 C      ARRAYS:
00130 C          TC=12-FOLD CDP-GATHER ON ENTRY AND WITH
00140 C          GROUND NOISE ADDED ON RETURN.
00150 C          X =OFFSET**2
00160 C
00170 C      YEN MAR-1984
00180 C*****
00190 C
00200 C      SUBROUTINE GROUND(TC,X)
00210 C      DIMENSION TC(700,1),X(1)
00220 C      DO 1 J=1,12
00230 C      XX=X(J)**0.5
00240 C      DO 2 K=1,4
00250 C*****LOCATE THE GROUND NOISE ARRIVING TIME
00260 C      NN=XX/(350.+K*50.)*500.0
00270 C      IF(NN.GT.700) GOTO 2
00280 C*****ADD GROUND NOISE WITH STRENGTH OF 0.35
00290 C      TC(NN,J)=TC(NN,J)+0.35
00300 C      2 CONTINUE
00310 C      1 CONTINUE
00320 C      RETURN
00330 C      END
*EU

```

```

00010 C*****
00020 C
00030 C      SUBROUTINE MODA
00040 C
00050 C*****
00060 C
00070 C      SETS THE ARRIVING TIME AND THE REFLECTION
00080 C-----COEFF FOR 12-FOLD CDP GATHER TRACES FROM THE
00090 C      THREE-LAYER MODEL AS FOLLOWS:
00100 C      LAYER 1:  WATERBOTTOM,  V=1500,  VRMS=1500,  Z=150.
00110 C      LAYER 2:  BOUNDARY A,   V=2000,  VRMS=1795,  Z=400.
00120 C      LAYER 3:  BOUNDARY B,   V=3000,  VRMS=2437,  Z=1000.
00130 C
00140 C      VARIABLES:
00150 C      TX=ARRIVING TIME OF OFFSET X FOR WATERBOT.
00160 C      TZ=ZERO OFFSET ARRIVING TIME FOR WATERBOT.
00170 C      TXA=ARRIVING TIME FOR BOUNDARY A.
00180 C      TXB=ARRIVING TIME FOR BOUNDARY B.
00190 C      ARRAYS:
00200 C      TC=12-FOLD CDP-GATHER.
00210 C      X =(OFFSET DISTANCE)**2 FOR 12 STATIONS.
00220 C      XV=(OFFSET/VRMS)**2.
00230 C
00240 C      YEN MAR-1984
00250 C*****
00260 C      SUBROUTINE MODA(TC,X,XV)
00270 C      DIMENSION TC(700,1),X(1),XV(1)
00280 C      DO 10 I=1,700
00290 C      DO 20 L=1,12
00300 C      20  TC(I,L)=0.0
00310 C      10  CONTINUE
00320 C*****SET OFFSET AND XV
00330 C      DO 30 I=1,12
00340 C      X(I)=(100.0+(I-1)*50.0)**2
00350 C      30  XV(I)=X(I)/2250000.0
00360 C*****SET ARRIVING TIME FOR WATE LAYER
00370 C      DO 40 IM=1,6
00380 C      TZ=(IM*0.2)**2
00390 C      DO 50 L=1,12
00400 C      TX=SQRT(TZ+XV(L))
00410 C      I=TX*500.0+1
00420 C      IF(I.GT.700) I=700
00430 C      50  TC(I,L)=-1.0*(-1.0)**IM
00440 C      40  CONTINUE
00450 C*****SET ARRIVING TIME FOR PRIMARIES
00460 C      DO 60 I=1,12
00470 C      XVA=X(I)/3222025.0
00480 C      XVB=X(I)/5938969.0
00490 C      TXA=SQRT(0.2025+XVA)
00500 C      TXB=SQRT(0.7225+XVB)
00510 C      IA=TXA*500.0+1
00520 C      IB=TXB*500.0+1
00530 C      TC(IA,I)=0.1+TC(IA,I)
00540 C      99  FORMAT(1X,'MODA DONE')
00550 C      60  TC(IB,I)=TC(IB,I)+0.2
00560 C      TYPE 99
00570 C      RETURN
00580 C      END
*EU

```

```

00010 C*****
00020 C
00030 C      SUBROUTINE MODB
00040 C
00050 C*****
00060 C
00070 C      SETS THE ARRIVING TIME AND THE REFLECTION
00080 C      COEFF FOR THE 4-LAYER MODEL AS FOLLOWS:
00090 C      LAYER 1: V=1500, VRMS=1500, Z=157, TZ=0.21
00100 C      LAYER 2: V=1700, VRMS=1618, Z=247, TZ=0.50
00110 C      LAYER 3: V=2090, VRMS=1843, Z=418, TZ=0.90
00120 C      LAYER 4: V=2555, VRMS=2142, Z=703, TZ=1.45
00130 C      REFLEC. COEFF. RESPECTIVELY:
00140 C      R12=0.6, R23=0.1, R34=0.3, R45=0.1
00150 C
00160 C      ARRAYS:
00170 C      TC=12-FOLD CDP-GATHER.
00180 C      X =OFFSET**2.
00190 C      XV=X/VRMS**2.
00200 C
00210 C      YEN MAR-1984
00220 C*****
00230 C      SUBROUTINE MODB(TC,X,XV)
00240 C      DIMENSION TC(700,1),X(1),XV(1)
00250 C*****SET OFFSET AND XV.
00260 C      DO 30 I=1,12
00270 C      X(I)=(200.0+(I-1)*100.0)**2
00280 C      30 XV(I)=X(I)/2250000.0
00290 C*****SET ARRIVING TIME OF WATER LAYER.
00300 C      DO 40 IM=1,6
00310 C      TZ=(IM*0.21)**2
00320 C      DO 50 L=1,12
00330 C      TX=SQRT(TZ+XV(L))
00340 C      NBEG=TX*500.0+1
00350 C      IF(NBEG.GT.700) GOTO 50
00360 C      TC(NBEG,L)=TC(NBEG,L)+0.6
00370 C      50 CONTINUE
00380 C      40 CONTINUE
00390 C*****SET ARRIVING TIME FOR PRIMARIES.
00400 C      DO 60 I=1,12
00410 C      XVA=X(I)/1618**2
00420 C      XVB=X(I)/1843.0**2
00430 C      XVC=X(I)/2142.0**2
00440 C      TXA=SQRT(0.25+XVA)*500.0
00450 C      TXB=SQRT(0.81+XVB)*500.0
00460 C      TVC=SQRT(2.1025+XVC)*500.0
00470 C      NA=TXA+1
00480 C      NB=TXB+1
00490 C      NC=TXC+1
00500 C      IF(NA.GT.700) GOTO 60
00510 C      IF(NB.GT.700) GOTO 60
00520 C      IF(NC.GT.700) GOTO 60
00530 C      TC(NA,I)=TC(NA,I)+0.064
00540 C      TC(NB,I)=TC(NB,I)+0.19
00550 C      TC(NC,I)=TC(NC,I)+0.058
00560 C      60 CONTINUE
00570 C      RETURN
00580 C      END
*EU

```

```

00010 *****
00020 C
00030 C      SUBROUTINE MODX
00040 C
00050 C*****
00060 C
00070 C      SETS A MODEL OF 6 CDP-GATHER TRACES.
00080 C
00090 C      VARIABLES:
00100 C          TNGS=TOTAL ENERGY OF SIGNAL.
00110 C          TNGN=TOTAL ENERGY OF NOISE.
00120 C
00130 C      ARRAYS:
00140 C          TC=6-FOLD CDP-GATHER AND ITS STACKS.
00150 C          BUF=BUFFER OF TC.
00160 C          SW=WAVELET.
00170 C          RATE=SIGNAL-TO-NOISE ENERGY RATIO.
00180 C          SCL=SIGNAL SCALE.
00190 C
00200 C      YEN MAR-1984
00210 C*****
00220 C      SUBROUTINE MODX(TC,BUF,SW,RATE,SCL)
00230 C          DIMENSION TC(280,7),BUF(280,6),SW(40)
00240 C          DIMENSION RATE(7),SCL(7)
00250 C*****SET RANDOM NOISE INTO TC & BUF.
00260 C          TNGS=4.0
00270 C          DO 1 J=1,6
00280 C              TNGN=0.0
00290 C              DO 2 I=1,280
00300 C                  YYY=(RAN(N)-0.5)*0.3
00310 C                  TNGN=YYY*YYY+TNGN
00320 C                  BUF(I,J)=YYY
00330 C              2 TC(I,J)=YYY
00340 C          1 RATE(J)=TNGS/TNGN
00350 C*****SET SIGNAL INTO TC & BUF.
00360 C          DO 3 J=1,7
00370 C              DO 4 KK=1,4
00380 C                  K=20+(KK-1)*70
00390 C                  BUF(K,J)=(BUF(K,J)+1.0)*SCL(J)
00400 C              4 TC(K,J)=BUF(K,J)
00410 C          3 CONTINUE
00420 C*****PRODUCE SYNTHETIC TRACES.
00430 C          CALL CONVTC(TC,SW,280,7,40)
00440 C          DO 5 J=1,6
00450 C              DO 6 I=1,280
00460 C          6 BUF(I,J)=TC(I,J)
00470 C          5 CONTINUE
00480 C          RETURN
00490 C          END
*EU

```

```

00010 C*****
00020 C
00030 C      SUBROUTINE MXSET
00040 C
00050 C*****
00060 C
00070 C      SETS A MATRIX OF AUTOCORRE X FOR THE
00080 C      DESIGN OF THE WIENER OPERATOR.
00090 C
00100 C      VARIABLES:
00110 C          N =DESIRED DIMENSION OF THE MATRIX.
00120 C          LDA=THE MAX MATRIX DIMENSION IN MAIN PROGRAM.
00130 C
00140 C      ARRAYS:
00150 C          A =RESULTED MATRIX ON RETURN.
00160 C          XX =AUTOCORR X ON ENTRY.
00170 C          BUFF=BUFFER TRACE.
00180 C
00190 C      YEN MAR-1984
00200 C*****
00210 C      SUBROUTINE MXSET(XX,N,A,LDA)
00220 C      DIMENSION A(LDA,1),XX(1),BUFF(120)
00230 C*****SET THE FIRST ROW
00240 C      DO 1 J=1,N
00250 C          BUFF(J)=XX(J)
00260 C      1  A(1,J)=XX(J)
00270 C*****SET THE REST ROWS
00280 C      DO 2 I=2,N
00290 C          DO 3 J=2,N
00300 C              K=N-J+2
00310 C      3  BUFF(K)=BUFF(K-1)
00320 C          BUFF(1)=XX(I)
00330 C          DO 4 J=1,N
00340 C      4  A(I,J)=BUFF(J)
00350 C      2  CONTINUE
00360 C          RETURN
00370 C      END
*EU

```

```

00010 C*****
00020 C
00030 C      SUBROUTINE NMO
00040 C
00050 C*****
00060 C
00070 C      MAKES NMO CORRECTION FOR THE ARRAY TC.
00080 C
00090 C      VARIABLES:
00100 C          TX=ARRIVING TIME OF OFFSET X.
00110 C          TZ=ARRIVING TIME OF ZERO OFFSET.
00120 C      ARRAYS:
00130 C          TC=12-FOLD CDP-GATHER ON ENTRY, AND WILL BE
00140 C          RETURN AFTER NMO CORRECTION.
00150 C          NA=VELOCITY DATA OF BOUNDARY A FROM KNOWN MODEL.
00160 C          NB=VELOCITY DATA OF BOUNDARY B FROM KNOWN MODEL.
00170 C          X=(OFFSET DISTANCE FOR EACH STATION)**2
00180 C          V=RMS VELOCITY FOR EACH DATA POINT.
00190 C
00200 C      YEN MAR-1984
00210 C*****
00220 C      SUBROUTINE NMO(TC,V,X)
00230 C          DIMENSION TC(700,1),V(1),X(1),NA(12),NB(12)
00240 C          DATA NA/116,117,118,119,121,122,124,126,127,129,131,132/
00250 C          DATA NB/241,241,241,241,241,242,242,243,243,243,244,244/
00260 C          DO 5 I=1,12
00270 C              5      X(I)=(100.0+(I-1)*50.0)**2
00280 C          DO 50 J=1,12
00290 C              N1=NA(J)
00300 C              N2=NB(J)
00310 C      SET RMS VELOCITY DOWN TO WATERBOTTOM.
00320 C          DO 10 I=1,N1
00330 C              10      V(I)=2250000.0
00340 C              DT=N1*0.002
00350 C              AA=V(N1)*DT
00360 C              KK=N1+1
00370 C      SET RMS VELOCITY DOWN TO BOUNDARY A.
00380 C          DO 20 I=KK,N2
00390 C              K=I-N1
00400 C              20      V(I)=(AA+K*8000.0)/(DT+K*0.002)
00410 C              DT=N2*0.002
00420 C              KK=N2+1
00430 C              AA=V(N2)*DT
00440 C      SET RMS VELOCITY DOWN TO BOUNDARY B.
00450 C          DO 30 I=KK,699
00460 C              K=I-N2
00470 C              30      V(I)=(AA+K*18000.0)/(DT+K*0.002)
00480 C              NPP=10
00490 C      EXEC NMO.
00500 C          DO 60 I=10,695
00510 C              TZ=(0.002*(I-1))**2
00520 C              TX=SQRT(TZ+X(J)/V(I))*500.0
00530 C              NTX=TX+1
00540 C              IF(NTX.GT.700) GOTO 60
00550 C              IF(NTX.LT.NPP) NTX=NPP
00560 C              TC(I,J)=TC(NTX,J)
00570 C              NPP=NTX
00580 C              60      CONTINUE
00590 C              50      CONTINUE
00600 C              RETURN
00610 C              END
*
```

```

00010 C*****
00020 C
00030 C      SUBROUTINE PLO
00040 C
00050 C*****
00060 C
00070 C      PLOTS 2-DIMENSIONAL ARRAY T.
00080 C
00090 C      VARIABLES:
00100 C          DX=SPACING BETWEEN TWO TRACES.
00110 C          NL=NUMBER OF LINES TO BE PLOTTED.
00120 C          NPT=NUMBER OF DATA POINTS IN ONE LINE.
00130 C          LDM=LEADING DIMENSION IN MAIN PROGRAM OF T.
00140 C          YGD=LOCATION IN Y-AXIS FOR THE LABEL.
00150 C          XSP=LOCATION IN X-AXIS FOR THE TRACE BEGGIN.
00160 C          FPN=LABEL CONTENT.
00170 C          YM =PLOT INCREMENT IN Y-AXIS.
00180 C          XM =PLOT INCREMENT IN X-AXIS.
00190 C      ARRAYS:
00200 C          T = NL-FOLD CDP-GATHER TRACES.
00210 C
00220 C      YEN MAR-1984
00230 C*****
00240 C      SUBROUTINE PLO(T,DX,NL,NPT,LDM)
00250 C          DIMENSION T(LDM,1)
00260 C*****FIND MAX VALUE AND NORMALIZING.
00270 C          AMAX=-1000.0
00280 C          DO 40 J=1,NL
00290 C          DO 42 I=1,NPT
00300 C              TT=ABS(T(I,J))
00310 C          42 IF(TT.GT.AMAX) AMAX=TT
00320 C          40 CONTINUE
00330 C          DO 44 J=1,NL
00340 C          DO 46 I=1,NPT
00350 C          46 T(I,J)=T(I,J)*0.5/AMAX
00360 C          44 CONTINUE
00370 C          IERR=-2
00380 C          CALL PLOTS(IERR,0)
00390 C*****PLOT LABEL.
00400 C          CALL PLOT(7.0,8.0,-3)
00410 C          CALL PLOT(0.0,-7.0,+2)
00420 C          DO 50 I=1,15
00430 C          YGD=-0.5*I+0.5
00440 C          CALL PLOT(0.0,YGD,3)
00450 C          CALL PLOT(0.25,YGD,2)
00460 C          YGD=YGD-0.12
00470 C          FPN=0.1*I-0.1
00480 C          50 CALL NUMBER(0.3,YGD,0.16,FPN,0.0,+2)
00490 C*****PLOT NLINE TRACES.
00500 C          DO 52 J=1,NL
00510 C          XSP=-0.33-DX*J
00520 C          CALL PLOT(XSP,0.0,+3)
00530 C          YM=0.0
00540 C          XM=XSP
00550 C          DO 54 I=1,NPT
00560 C          YM=YM-0.01
00570 C          XM=XSP+T(I,J)
00580 C          54 CALL PLOT(XM,YM,+2)
00590 C          52 CONTINUE

```

```

00600      TYPE 99
00610      99  FORMAT(1X,'PLOT FINISHED')
00620      CALL PLOT(XM,YM,999)
00630      RETURN
00640      END

```

```

00010      C*****
00020      C
00030      C          SUBROUTINE PLOLAB
00040      C
00050      C*****
00060      C
00070      C          INITIALS THE PLOT AND PLOTS THE LABEL
00080      C          FOR 0-1.4 SEC TIME AXIX.
00090      C
00100      C          YEN MAR-1984
00110      C*****
00120      C          SUBROUTINE PLOLAB
00130      C          IERR=-2
00140      C          CALL PLOTS(IERR,0)
00150      C*****PLOT LABEL.
00160      C          CALL PLOT(16.0,9.,-3)
00170      C          CALL PLOT(0.0,-8.,+2)
00180      C          DO 50 I=1,17
00190      C          YGD=-0.5*I+0.5
00200      C          CALL PLOT(0.0,YGD,3)
00210      C          CALL PLOT(0.25,YGD,2)
00220      C          YGD=YGD-0.12
00230      C          FPN=0.1*I-0.1
00240      C          50 CALL NUMBER(0.3,YGD,0.16,FPN,0.0,+2)
00250      C          RETURN
00260      C          END
*EU

```



```

00010 C*****
00020 C
00030 C      SUBROUTINE PLOTX
00040 C
00050 C*****
00060 C
00070 C      PLOTS THE 11 MODX TRACES WITH RATES AND SCL.
00080 C
00090 C      ARRAYS:
00100 C          TC:  11 TRACES TO BE PLOTTED.
00110 C          RATE : S/N RATIO OF THE 6 MODEL TRACES.
00120 C          SCL  : SIGNAL SCALE OF THE 6 TRACES.
00130 C
00140 C      YEN MAR-1984
00150 C*****
00160 C      SUBROUTINE PLOTX(T,RATE,SCL)
00170 C      DIMENSION T(280,1),RATE(1),SCL(1)
00180 C*****NORMALIZATION OF THE DATA
00190 C      AMAX=-1000.0
00200 C      DO 40 J=1,11
00210 C      DO 42 I=1,280
00220 C      TT=ABS(T(I,J))
00230 C      42  IF(TT.GT.AMAX) AMAX=TT
00240 C      40  CONTINUE
00250 C      DO 44 J=1,11
00260 C      DO 46 I=1,280
00270 C      46  T(I,J)=T(I,J)*0.5/AMAX
00280 C      44  CONTINUE
00290 C*****START PLOTTING
00300 C      IERR=-2
00310 C      CALL PLOTS(IERR,0)
00320 C      CALL PLOT(7.0,8.0,-3)
00330 C      DO 52 J=1,11
00340 C      XSP=-0.33-0.33*J
00350 C      CALL PLOT(XSP,0.,+3)
00360 C      YM=0.0
00370 C      XM=XSP
00380 C      DO 54 I=1,280
00390 C      YM=YM-0.02
00400 C      XM=XSP+T(I,J)
00410 C      54  CALL PLOT(XM,YM,+2)
00420 C      IF(J .GT. 6) GOTO 52
00430 C      YM=YM-0.5
00440 C      CALL PLOT(XSP,YM,+3)
00450 C      FPN=RATE(J)
00460 C      CALL NUMBER(XSP,YM,0.1,FPN,90.0,2)
00470 C*****PLOT S/N RATE
00480 C      YM=YM-1.0
00490 C      FPN=SCL(J)
00500 C      CALL PLOT(XSP,YM,+3)
00510 C*****PLOT SIGNAL SCALE
00520 C      CALL NUMBER(XSP,YM,0.1,FPN,90.0,2)
00530 C      52  CONTINUE
00540 C      CALL PLOT(0,0,+999)
00550 C      RETURN
00560 C      END
*EU

```

```

00010 C*****
00020 C
00030 C      SUBROUTINE PLOZ
00040 C
00050 C*****
00060 C
00070 C      PLOTS ONE TRACE DATA.
00080 C
00090 C      VARIABLES:
00100 C          DX=SPACING BETWEEN TWO TRACES.
00110 C          J =THE NUMBER OF THE TRACE IN MAIN PROGRAM.
00120 C      ARRAYS:
00130 C          T=TRACE TO BE PLOTED.
00140 C
00150 C      YEN MAR-1984
00160 C*****
00170 C      SUBROUTINE PLOZ(T,DX,J)
00180 C          DIMENSION T(1)
00190 C*****NORMALIZATION
00200 C          AMAX=50.0
00210 C          DO 42 I=1,700
00220 C          TT=ABS(T(I))
00230 C          IF(TT.GT.AMAX) AMAX=TT
00240 C      42 CONTINUE
00250 C*****START PLOTTING
00260 C          DO 46 I=1,700
00270 C      46 T(I)=T(I)*0.7/AMAX
00280 C          XSP=-.33-DX*J
00290 C          CALL PLOT(XSP,0.,+3)
00300 C          YM=0.0
00310 C          XM=XSP
00320 C          DO 54 I=1,700
00330 C          YM=YM-0.01
00340 C          XM=XSP+T(I)
00350 C      54 CALL PLOT(XM,YM,+2)
00360 C          RETURN
00370 C          END
*EU

```

```
00010 C*****
00020 C
00030 C      SUBROUTINE PRDZX
00040 C
00050 C*****
00060 C
00070 C      SETS DESIRED OUTPUT FOR PREDICTIVE DECON.
00080 C
00090 C      VARIABLES:
00100 C          NX=NUMBER OF DATA POINTS IN THE TRACE WINDOW.
00110 C          NR=PREDICTIVE LENGTH.
00120 C      ARRAYS:
00130 C          XX=TRACE ON ENTRY.
00140 C          ZX=TRACE OF PREDICTIVE OUTPUT ON RETURN.
00150 C
00160 C      YEN MAR-1984
00170 C*****
00180 C      SUBROUTINE PRDZX(XX,ZX,NX,NR)
00190 C      DIMENSION XX(1),ZX(1)
00200 C      DO 1 I=1,NX
00210 C          II=NR+I
00220 C      1  ZX(I)=XX(II)
00230 C      RETURN
00240 C      END
*EU
```

```

00010 C*****
00020 C
00030 C      SUBROUTINE RAM
00040 C
00050 C*****
00060 C
00070 C      GENERATES RANDOM VALUES INTO ARRAY TC
00080 C      AND WEIGHTED RANDOM VALUES BY FACT.
00090 C
00100 C      VARIABLES:
00110 C          SUMS=SUM OF SIGNAL AMPLITUDES.
00120 C          SUMN=SUM OF RANDOM NOISE AMPLITUDES.
00130 C          YYY=RANDOM VALUE GENERATED FROM RAN(N)
00140 C          FUNCTION, VALUE RANGES FROM -.5 TO .5.
00150 C          SN =SIGNAL-TON-NOISE AMP RATIO
00160 C          FACT=WEIGHTED VALUE, IT WAS 0.06 BEFORE.
00170 C      ARRAYS:
00180 C          TC =12-FOLD CDP TRACES.
00190 C
00200 C      YEN MAR-1984
00210 C*****
00220 C      SUBROUTINE RAM(TC,FACT)
00230 C      DIMENSION TC(700,1)
00240 C*****SET SUM OF TOTAL SIGNAL ENERGY
00250 C      SUMS=0.01*12.
00260 C      SUMN=0.0
00270 C*****SET RANDOM VALUES AND GET ITS SUM
00280 C      DO 10 I=50,690
00290 C      DO 20 J=1,12
00300 C      YYY=RAN(N)-0.5
00310 C      SUMN=SUMN+(YYY*FACT)**2
00320 C      20 TC(I,J)=TC(I,J)+YYY*FACT
00330 C      10 CONTINUE
00340 C      SN=SUMS/(SUMN/640)
00350 C*****COMPUTE THE S/N AMP RATIO
00360 C      SN=SQRT(SN)
00370 C      TYPE 100,SN
00380 C      TYPE 99
00390 C      99 FORMAT(1X,'RAM DONE')
00400 C      100 FORMAT(1X,'AMPS/N=',F11.3)
00410 C      RETURN
00420 C      END
00430
*EU

```

```

00010 C*****
00020 C
00030 C      SUBROUTINE RICKER
00040 C
00050 C*****
00060 C
00070 C      GENERATES A RICKER WAVELET.
00080 C
00090 C      VARIABLES:
00100 C          DT=TIME STEP BETWEEN DATA POINTS.
00110 C          F =CENTER FREQUENCY.
00120 C          NFP=NUMBER OF POINTS FOR WAVELET.
00130 C          T=TIME IN SECONDS.
00140 C      ARRAYS:
00150 C          FILT=WAVELET.
00160 C
00170 C      YEN MAR-1984
00180 C*****
00190 C      SUBROUTINE RICKER(F,DT,FILT,NFP)
00200 C          DIMENSION FILT(1)
00210 C          PI=3.14159
00220 C          T=0.0
00230 C          FACT=PI*F
00240 C          I1=(NFP+1)/2
00250 C*****COMPUTE THE RIGHT-HALF WAVELET
00260 C          DO 1 I=I1,NFP
00270 C              R1=FACT*T
00280 C              R1SQ=R1*R1
00290 C              R2=1-2.*R1SQ
00300 C              R3=EXP(-R1SQ)
00310 C              FILT(I)=R2*R3*(-25.0)
00320 C          1 T=T+DT
00330 C*****COPY THE LEFT SYMMETRICAL PART
00340 C          I2=I1-1
00350 C          J=NFP
00360 C          DO 2 I=1,I2
00370 C              FILT(I)=FILT(J)
00380 C          2 J=J-1
00390 C          RETURN
00400 C      END
*
```

```
00010 C*****
00020 C
00030 C      SUBROUTINE SGEFA(A,LDA,N,IPVT,INFO)
00040 C
00050 C      SUBROUTINE SGESL(A,LDA,N,IPVT,B,JOB)
00060 C
00070 C*****
00080 C
00090 C      SUBROUTINES SOLVE LINEAR EQUATION AX=B.
00100 C      MAKE REFERENCE TO THE SOFTWARE LINPAC FOR CODES.
00110 C      WHEN EXEC USE LBY:LINPAC/LIBRARY.
00120 C
00130 C      ON ENTRY:
00140 C      A = A 2-D ARRAY WITH DIMENSION(LDA,N)
00150 C      LDA= THE LEADING DIMENSION OF THE MATRIX A.
00160 C      N = THE ORDER OF THE MATRIX A.
00170 C      JOB=IF 0, THE SYSTEM AX=B IS SOLVED.
00180 C      IF NON-ZERO, A(*)X=B IS SOLVED.
00190 C
00200 C      ON RETURN:
00210 C      A = CONTAINS IN ITS UPPER TRIANGLE AN
00220 C      UPPER TRIANGLE MATRIX.
00230 C      IPVT= INTEGER ARRAY OF DIMENSION N WITH PIVOT
00240 C      INFORMATION.
00250 C      INFO= AN INTEGER ARRAY, IF 0 THE PROCESS IS OK.
00260 C      THE VALUE OF INFO SHOULD BE CHECKED.
00270 C      B = CONTAINS THE SOLUTION, X.
00280 C
00290 C      AFTER LINPAC 1979
00300 C*****
*
```

```

00010 C*****
00020 C
00030 C      SUBROUTINE SINWAV
00040 C
00050 C*****
00060 C
00070 C      GENERATES A WEIGHTED SINWAVE.
00080 C
00090 C      VARIABLES:
00100 C      DT=TIME STEP BETWEEN DATA POINTS IN SEC.
00110 C      F =FREQUENCY OF SINWAVE.
00120 C      NFP=NUMBER OF POINTS FOR WAVELET ON RETURN.
00130 C      NW=NUMBER OF POINTS FOR WAVELET PERIOD.
00140 C      PERD=WAVELET PERIOD IN SEC.
00150 C      ARRAY:
00160 C      SW=SINWAVE ON RETURN.
00170 C
00180 C      YEN MAR-1984
00190 C*****
00200 C      SUBROUTINE SINWAV(F,DT,SW,NFP)
00210 C      DIMENSION SW(1)
00220 C*****SET PERIOD AND THE DATA POINTS OF SINWAVE
00230 C      PERD=1.0/F
00240 C      NW=PERD/DT
00250 C      PI=3.1416
00260 C*****SET INCREMENT AND WEIGHT
00270 C      PSIN=(PI/180.)*(360./NW)
00280 C      PCOS=(PI/180.)*(90.0/NFP)
00290 C*****BUILD SINWAVE
00300 C      DO 10 I=1,NFP
00310 C      K=I-1
00320 C      GA=K*PSIN
00330 C      GB=K*PCOS
00340 C      SW(I)=SIN(GA)*COS(GB)
00350 C      10 CONTINUE
00360 C      RETURN
00370 C      END
*

00010 C*****
00020 C
00030 C      SUBROUTINE SPECPL
00040 C
00050 C*****
00060 C
00070 C      PLOTS THE AMPLITUDE AND THE PHASE SPECTRUMS
00080 C      FOR THE FILTER.
00090 C
00100 C      VARIABLES:
00110 C      NFP=NUMBER OF DATA POINTS OF THE FILTER.
00120 C      DT =TIME STEP BETWEEN TWO DATA.

```

```

00130 C      ARRAYS:
00140 C      A =COMPLEX ARRAY FOR FFT PROCESS.
00150 C      IT IS THE FILTER ON ENTRY,
00160 C      AND IT WILL BE THE FREQ -DOMAIN ON RETURN.
00170 C      FREQ=FREQUENCY VALUE FOR PLOTTING PAIR.
00180 C      AMP =AMPLITUDE VALUE FOR PLOTTING PAIR.
00190 C      PHZ =PHASE VALUE FOR PLOTTING PAIR.
00200 C
00210 C      YEN MAR-1984
00220 C*****
00230 C      SUBROUTINE SPECPL(NFF,FILT,DT)
00240 C      COMPLEX A(1024)
00250 C      DIMENSION FILT(1),FREQ(514),AMP(514),PHZ(514)
00260 C*****FILL THE FILTER INTO A
00270 C      DO 1 I=1,NFF
00280 C      T=FILT(I)
00290 C      A(I)=CMPLX(T,0.0)
00300 C      1 CONTINUE
00310 C*****TRANSFORM
00320 C      CALL FFT(1024,A,1.0)
00330 C*****SET NYQUIST FREQUENCY
00340 C      ADD=1./(512*2*DT)
00350 C      FREQ(1)=0.0
00360 C      DO 2 I=2,512
00370 C      FREQ(I)=FREQ(I-1)+ADD
00380 C*****SET PHASE AND NORMALIZE AMP
00390 C      AMPMAX=-1E5
00400 C      DO 3 I=1,512
00410 C      AMP(I)=(REAL(A(I))*2+AIMAG(A(I))*2)**0.5
00420 C      IF(AMP(I).GT.AMPMAX) AMPMAX=AMP(I)
00430 C      3 PHZ(I)=ATAN2(AIMAG(A(I)),REAL(A(I)))
00440 C*****MAKE THE PHASE CONTINUOUS
00450 C      PJ=0.0
00460 C      DO 4 I=2,512
00470 C      IF(ABS(PHZ(I)+PJ-PHZ(I-1))-3.141593) 40,40,10
00480 C      10 IF(PHZ(I)+PJ-PHZ(I-1)) 20,40,30
00490 C      20 PJ=PJ+3.141593*2.
00500 C      GOTO 40
00510 C      30 PJ=PJ-3.141593*2.
00520 C      40 PHZ(I)=PHZ(I)+PJ
00530 C      DO 4 I=1,512
00540 C      4 AMP(I)=AMP(I)/AMPMAX
00550 C*****PLOT (FREQ VS. AMP) &(FREQ VS. PHZ)
00560 C      IERR=-2
00570 C      CALL PLOTS(IERR,0)
00580 C      CALL SCALE(AMP,3.,512)
00590 C      CALL SCALE(PHZ,3.,512)
00600 C      CALL SCALE(FREQ,4.,512)
00610 C      CALL AXIS(1.,5.,9HFREQUENCY,-9,4.,0.,FREQ(513),FREQ(514))
00620 C      CALL AXIS(1.,5.,9HAMPLITUDE,+9,3.,90.,AMP(513),AMP(514))
00630 C      CALL AXIS(1.,1.,9HFREQUENCY,-9,4.,0.,FREQ(513),FREQ(514))
00640 C      CALL AXIS(1.,1.,5HPHASE,+5,3.,90.,PHZ(513),PHZ(514))
00650 C      CALL PLOT(1.,1.,-3)
00660 C      CALL LINE(FREQ,PHZ,512,0,0,0)
00670 C      CALL PLOT(0.,4.,-3)
00680 C      CALL LINE(FREQ,AMP,512,0,0,0)
00690 C      CALL PLOT(0,0,+999)
00700 C      RETURN
00710 C      END
*
```



```

00010 C*****
00020 C
00030 C      SUBROUTINE STK
00040 C
00050 C*****
00060 C
00070 C      PROCESSES ITERATIVE STACK ON TC.
00080 C
00090 C      VARIABLES:
00100 C          NN=NUMBER OF TRACES FOR STK DECIDED BY MUTING.
00110 C          TPOS=SUM OF THE POSITIVE GROUP.
00120 C          TNEG=SUM OF THE NEGATIVE GROUP.
00130 C          JOB=0: ITERATIVE PROCESS BUT NOT STACK.
00140 C              1: SINGLE-TRACE-PROCESS AND REPLAE TC.
00150 C              2: ITERATIVE STACK AND REPLACE TC.
00160 C          NITER=ITERATIONS WHEN CHOOSE JOB=0.
00170 C              OTHERWISE IT IS 12.
00180 C      ARRAYS:
00190 C          TC=12-FOLD CDP-GATHER ON ENTRY.
00200 C          1-12TH ITERATION RESULT ON RETURN.
00210 C          X =BUFFER.
00220 C
00230 C      YEN MAR-1984
00240 C*****
00250 C      SUBROUTINE STK(TC,NITER,JOB)
00260 C      DIMENSION TC(700,1),X(12)
00270 C      DO 50 I=1,700
00280 C          GG=I
00290 C*****SET MUTING FACTOR
00300 C          NN=0.465*SQRT(GG)
00310 C          IF(NN.LT.2) NN=2
00320 C          IF(NN.GT.12) NN=12
00330 C*****PROCESS ITERATIVE STACK
00340 C          DO 60 ITER=1,NITER
00350 C              TPOS=0.0
00360 C              TNEG=0.0
00370 C              DO 70 J=1,NN
00380 C                  IF(TC(I,J) .GT. 0.0) TPOS=TPOS+TC(I,J)
00390 C              70 IF(TC(I,J) .LT. 0.0) TNEG=TNEG+TC(I,J)
00400 C                  TPOS=TPOS/NN
00410 C                  TNEG=TNEG/NN
00420 C              DO 75 J=1,12
00430 C                  IF(TC(I,J) .GT. TPOS) TC(I,J)=TPOS
00440 C                  IF(TC(I,J) .LT. TNEG) TC(I,J)=TNEG
00450 C              75 CONTINUE
00460 C*****SELECT ITERATIVE OUTPUT
00470 C          IF(JOB.EQ.2) X(ITER)=TPOS+TNEG
00480 C          IF(JOB.EQ.1) X(ITER)=TC(I,1)
00490 C          60 CONTINUE
00500 C          IF(JOB.EQ.0) GOTO 50
00510 C*****TRANSFER THE WANTED OUTPUT TO TC
00520 C          DO 80 K=1,NITER
00530 C              80 TC(I,K)=X(K)
00540 C          50 CONTINUE
00550 C          RETURN
00560 C          END
*
```

```

00010 C*****
00020 C
00030 C      SUBROUTINE STKGEN
00040 C
00050 C*****
00060 C
00070 C      PROCESSES ITERATIVE STACK ON TC.
00080 C      THIS IS A GENERALIZED FORM WITHOUT MUTING.
00090 C
00100 C      VARIABLES:
00110 C          NN=NUMBER OF TRACES FOR STACKING.
00120 C          NPT=DATA POINTS OF ONE TRACE.
00130 C          TPOS=SUM OF THE POSITIVE GROUP.
00140 C          TNEG=SUM OF THE NEGATIVE GROUP.
00150 C          JOB=0: ITERATIVE PROCESS BUT NOT STACK.
00160 C              1: SINGLE-TRACE-PROCESS AND REPLAE TC.
00170 C              2: ITERATIVE STACK AND REPLACE TC.
00180 C          NITER=ITERATIONS WHEN CHOOSE JOB=0.
00190 C              OTHERWISE IT IS NN.
00200 C      ARRAYS:
00210 C          TC=NN-FOLD CDP-GATHER ON ENTRY.
00220 C          1-NN ITERATION RESULT ON RETURN.
00230 C          X =BUFFER.
00240 C
00250 C      YEN MAR-1984
00260 C*****
00270 C      SUBROUTINE STKGEN(TC,NPT,NN,NITER,JOB)
00280 C      DIMENSION TC(NPT,1),X(24)
00290 C*****PROCESS ITERATIVE STACK
00300 C      DO 50 I=1,NPT
00310 C      DO 60 ITER=1,NITER
00320 C          TPOS=0.0
00330 C          TNEG=0.0
00340 C          DO 70 J=1,NN
00350 C              IF(TC(I,J) .GT. 0.0) TPOS=TPOS+TC(I,J)
00360 C          70 IF(TC(I,J) .LT. 0.0) TNEG=TNEG+TC(I,J)
00370 C              TPOS=TPOS/NN
00380 C              TNEG=TNEG/NN
00390 C          DO 75 J=1,NN
00400 C              IF(TC(I,J) .GT. TPOS) TC(I,J)=TPOS
00410 C              IF(TC(I,J) .LT. TNEG) TC(I,J)=TNEG
00420 C          75 CONTINUE
00430 C*****SELECT OUTPUT
00440 C          IF(JOB.EQ.2) X(ITER)=TPOS+TNEG
00450 C          IF(JOB.EQ.1) X(ITER)=TC(I,1)
00460 C          60 CONTINUE
00470 C          IF(JOB.EQ.0) GOTO 50
00480 C*****TRANSFER WANTED OUTPUT TO TC
00490 C          DO 80 K=1,NITER
00500 C          80 TC(I,K)=X(K)
00510 C          50 CONTINUE
00520 C          RETURN
00530 C          END
*
```

```

00010 C*****
00020 C
00030 C      SUBROUTINE STKLEV
00040 C
00050 C*****
00060 C
00070 C      STKLEV MAKES ITERATIVE STACK ON ONE LEVEL DATA.
00080 C
00090 C      VARIABLES:
00100 C          N=DATA POINTS OF ONE LEVEL.
00110 C          NITER=ITERATIONS TO BE RUN WHEN JOB=0.
00120 C          JOB=0: ITERATIVE PROCESS WITHOUT REPLACING.
00130 C          =1: SINGLE-TRACE-PROCESS AND REPLACING.
00140 C          =2: ITERATIVE STACKING AND REPLACING.
00150 C      ARRAYS:
00160 C          X=ONE LEVEL DATAB FROM TRACE 1 TO N.
00170 C          BUF=BUFFER FOR STORING THE RESULT STACK.
00180 C
00190 C      YEN MAR-1984
00200 C*****
00210 C      SUBROUTINE STKLEV(X,N,NITER,JOB)
00220 C      DIMENSION X(1),BUF(12)
00230 C*****PROCESS ITERATIVE STACK
00240 C      DO 60 ITER=1,NITER
00250 C      TPOS=0.0
00260 C      TNEG=0.0
00270 C      DO 70 J=1,N
00280 C      IF(X(J).GT. 0.0) TPOS=TPOS+X(J)
00290 C      70 IF(X(J).LT. 0.0) TNEG=TNEG+X(J)
00300 C      TPOS=TPOS/N
00310 C      TNEG=TNEG/N
00320 C      DO 75 J=1,N
00330 C      IF(X(J).GT.TPOS) X(J)=TPOS
00340 C      75 IF(X(J) .LT.TNEG) X(J)=TNEG
00350 C*****SELECT OUTPUT
00360 C      IF(JOB.EQ.2) BUF(ITER)=TPOS+TNEG
00370 C      IF(JOB.EQ.1) BUF(ITER)=X(1)
00380 C      60 CONTINUE
00390 C      IF(JOB.EQ.0) GOTO 99
00400 C      DO 80 I=1,N
00410 C*****TRANSFER WANTED OUTPUT TO TC
00420 C      80 X(I)=BUF(I)
00430 C      99 CONTINUE
00440 C      RETURN
00450 C      END
*
```

```

00010 C*****
00020 C
00030 C      SUBROUTINE TRACE
00040 C
00050 C*****
00060 C
00070 C      PLOTS A TIME SERIES.
00080 C
00090 C      VARIABLES:
00100 C          NFP=DATA POINTS OF THE TIME TRACE.
00110 C          DT =TIME STEP BETWEEN TWO DATA.
00120 C      ARRAY:
00130 C          FREQ=TIME AXIX WITH INCREMENT OF DT.
00140 C          AMP=THE AMPLITUDE OF THE INPUT FILT.
00150 C
00160 C      YEN MAR-1984
00170 C*****
00180 C      SUBROUTINE TRACE(NFP,AMP,DT)
00190 C      DIMENSION AMP(702),FREQ(702)
00200 C*****SET TIME WINDOW
00210 C      DO 2 I=1,NFP
00220 C      2      FREQ(I)=(I-1)*DT
00230 C*****PLOT (FREQ VS. AMP)
00240 C      IERR=-2
00250 C      CALL PLOTS(IERR,0)
00260 C      CALL SCALE(AMP,3.,NFP)
00270 C      CALL SCALE(FREQ,5.,NFP)
00280 C      NA=NFP+1
00290 C      NB=NFP+2
00300 C      CALL AXIS(1.,1.,4HTIME,-4,5.,0.,FREQ(NA),FREQ(NB))
00310 C      CALL AXIS(1.,1.,9HAMPLITUDE,+9,3.,90.,AMP(NA),AMP(NB))
00320 C      CALL PLOT(1.,1.,-3)
00330 C      CALL LINE(FREQ,AMP,NFP,0,0,0)
00340 C      CALL PLOT(0,0,+999)
00350 C      RETURN
00360 C      END
*
```

```

00010 C*****
00020 C
00030 C      SUBROUTINE TRSET
00040 C
00050 C*****
00060 C
00070 C      SES A WINDOW TRACE FOR THE WIENER PROCESS.
00080 C      THE WINDOW BEGINS AT THE FIRST REFLECTION,
00090 C      AND ENDS AT POINT 650.
00100 C
00110 C      VARIABLES:
00120 C      L =OFFSET LINE NUMBER ON ENTRY.
00130 C      NST=STARTING POINT OF THE FIRST REFLECTION.
00140 C      NW =WINDOW LENGTH.
00150 C      NZ = IF TRACE WAS NMO-CORRECTED NZ IS 0.
00160 C      XV=(OFFSET/VELOCITY)**2.
00170 C      ARRAYS:
00180 C      X =INPUT TRACE ON ENTRY,
00190 C      AND THE WINDOW TRACE ON RETURN,
00200 C      DATA AFTER NW ARE ZEROES.
00210 C
00220 C      YEN MAR-1984
00230 C*****
00240 C      SUBROUTINE TRSET(X,L,NST,NW,NZ)
00250 C      DIMENSION X(1)
00260 C*****FIND THE START POINT OF THE 1ST REFLECTION
00270 C      XV=(100.0+(L-1)*50.0)**2/2250000.0
00280 C      NST=1+500.0*SQRT(0.04+XV)
00290 C*****FIND THE WINDOW LENGTH
00300 C      NW=650-NST
00310 C      IF(NZ.EQ.0) NST=101
00320 C*****RESET THE INPUT TRACE
00330 C      DO 2 K=1,NW
00340 C      KK=K+NST
00350 C      2   X(K)=X(KK)
00360 C      KK=NW+1
00370 C*****ZERO THE TAIL
00380 C      DO 3 K=KK,700
00390 C      3   X(K)=0.0
00400 C      RETURN
00410 C      END
*
```

```
00010 C*****
00020 C
00030 C      SUBROUTINE WTD
00040 C
00050 C*****
00060 C
00070 C      MAKES OPTIMUM WEIGHTING ON TC.
00080 C
00090 C      VARIABLES:
00100 C      WT=OPTIMUM WEIGHTED RATIO.
00110 C      ARRAYS:
00120 C      TC=6-FOLD CDP-GATHER.
00130 C      BUF=BUFFER OF TC FOR WEIGHTED PROCESS.
00140 C      RATE=S/N ENERGY RATIO.
00150 C      SCL=SIGNAL SCALE.
00160 C
00170 C      YEN MAR-1984
00180 C*****
00190 C      SUBROUTINE WTD(TC,BUF,RATE,SCL)
00200 C      DIMENSION TC(280,1),BUF(280,1),RATE(1),SCL(1)
00210 C      DO 1 J=1,6
00220 C      QQ=RATE(J)
00230 C      WT=SQRT(QQ)/SCL(J)
00240 C      DO 2 I=1,280
00250 C      2  BUF(I,J)=TC(I,J)*WT
00260 C      1  CONTINUE
00270 C      RETURN
00280 C      END
*
```

APPENDIX C

Output of the Program 'GRAY'

THIS GENRAY FILE WAS CREATED 26-Jan-84
AT 26:08 HOURS

YOUR SOURCE-RECEIVER SECTION IS AS FOLLOWS:

```

MAX. NO. OF LEGS, NLEGS = R
NO. OF LAYERS ABOVE LOWER HALFSPACE, NLAY = 3
SOURCE 0=1 SV=-1 SH=0 SOUR = 1
CONVECTIONS, REFL. ONLY=1 NONE=0 ALL=-1. CONV = 1
SOURCE DEPTH, HS = 0.100000E-01
FIRST RECEIVER, HR = 0.100000E-01 R = 100.0000
NR=12 RECEIVER# SEPARATED BY DR = 50.00000 DZ = 0.000

```

THE EARTH-MODEL YOU HAVE CONSTRUCTED LOOKS LIKE:

LAYER	VP	VS	DM	ZI
1	1.500	0.100	1.000	150.000
2	2.000	1.100	2.000	250.000
3	3.000	1.600	3.000	600.000
4	3.500	1.940	2.500	99.990

THE DISPLAY SECTION YOU HAVE SELECTED IS AS FOLLOWS:

```

SAMPLE INTERVAL = 2.000 (DELT) MILLISECONDS
INITIAL TIME = 100.000 (MNMND) MILLISECONDS
FINAL TIME = 100.000 (MXMND) MILLISECONDS
SOURCE ENERGY = 1.000 (ENERGY) KILOJOULES.

```

```

THE MAXIMUM GAUSSIAN DISPLACEMENT ON A SPHERICAL SOURCE OF RADIUS
1.000 (SRAD) METERS IS .132E+04 MICRONS.
ALL PARTICLE DISPLACEMENTS CALCULATED IN MICRONS.

```

```

SUMMARY OF ARRIVAL TIMES AND AMPLITUDES FOR EACH RAY NAME
POSITIVE GROUND MOTION IS UP AND AWAY FROM THE SOURCE.
DIR: DIRECTION TAKEN BY RAY FROM THE SOURCE, 1 MEANS UP AND
-1 MEANS DOWN
ND: NUMBER OF DYNAMIC ANALOGS (SEE ENGELKEFER (I-2175))
THE SOURCE DEPTH IS 0.010 AND THE SOURCE IS IN LAYER # 1

```

RECEIVER # 1 IN LAYER # 1 , RECEIVER POSI : R= 100.0000 HR= 0.3120

SPECIES ARRIVAL TIME VERTICAL COMPG. RADIAL COMP. DIR ND RAYNAME

Species	Arrival Time	Vertical Comp.	Radial Comp.	Dir	Nd	Rayname
1	0.21	0.8	0.0	0	0	P1P1
1	0.4	0.8	0.0	0	0	P1P1
1	0.6	0.8	0.0	0	0	P1P1
1	0.7	0.8	0.0	0	0	P1P1
1	0.8	0.8	0.0	0	0	P1P1
1	0.9	0.8	0.0	0	0	P1P1
1	1.0	0.8	0.0	0	0	P1P1
1	1.1	0.8	0.0	0	0	P1P1
1	1.2	0.8	0.0	0	0	P1P1
1	1.3	0.8	0.0	0	0	P1P1
1	1.4	0.8	0.0	0	0	P1P1
1	1.5	0.8	0.0	0	0	P1P1
1	1.6	0.8	0.0	0	0	P1P1
1	1.7	0.8	0.0	0	0	P1P1
1	1.8	0.8	0.0	0	0	P1P1
1	1.9	0.8	0.0	0	0	P1P1
1	2.0	0.8	0.0	0	0	P1P1
1	2.1	0.8	0.0	0	0	P1P1
1	2.2	0.8	0.0	0	0	P1P1
1	2.3	0.8	0.0	0	0	P1P1
1	2.4	0.8	0.0	0	0	P1P1
1	2.5	0.8	0.0	0	0	P1P1
1	2.6	0.8	0.0	0	0	P1P1
1	2.7	0.8	0.0	0	0	P1P1
1	2.8	0.8	0.0	0	0	P1P1
1	2.9	0.8	0.0	0	0	P1P1
1	3.0	0.8	0.0	0	0	P1P1
1	3.1	0.8	0.0	0	0	P1P1
1	3.2	0.8	0.0	0	0	P1P1
1	3.3	0.8	0.0	0	0	P1P1
1	3.4	0.8	0.0	0	0	P1P1
1	3.5	0.8	0.0	0	0	P1P1
1	3.6	0.8	0.0	0	0	P1P1
1	3.7	0.8	0.0	0	0	P1P1
1	3.8	0.8	0.0	0	0	P1P1
1	3.9	0.8	0.0	0	0	P1P1
1	4.0	0.8	0.0	0	0	P1P1
1	4.1	0.8	0.0	0	0	P1P1
1	4.2	0.8	0.0	0	0	P1P1
1	4.3	0.8	0.0	0	0	P1P1
1	4.4	0.8	0.0	0	0	P1P1
1	4.5	0.8	0.0	0	0	P1P1
1	4.6	0.8	0.0	0	0	P1P1
1	4.7	0.8	0.0	0	0	P1P1
1	4.8	0.8	0.0	0	0	P1P1
1	4.9	0.8	0.0	0	0	P1P1
1	5.0	0.8	0.0	0	0	P1P1
1	5.1	0.8	0.0	0	0	P1P1
1	5.2	0.8	0.0	0	0	P1P1
1	5.3	0.8	0.0	0	0	P1P1
1	5.4	0.8	0.0	0	0	P1P1
1	5.5	0.8	0.0	0	0	P1P1
1	5.6	0.8	0.0	0	0	P1P1
1	5.7	0.8	0.0	0	0	P1P1
1	5.8	0.8	0.0	0	0	P1P1
1	5.9	0.8	0.0	0	0	P1P1
1	6.0	0.8	0.0	0	0	P1P1

