

EEE 407/591 PROJECT  
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# DATA COMPRESSION USING THE FFT

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## INTRODUCTION

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Data compression is one of the necessities of modern day. For instance, with the explosive growth of the Internet there is a growing need for audio compression, or data compression in general. One goal of such compressions is to minimize the storage space. Nowadays a 40GB hard drive can be bought within hundred dollars that makes the storage less of a problem. However, compression is greatly needed to reduce transmission bandwidth requirements, which can be achieved by data compression. Today all kind of audio/video is preferred in digital domain. Almost every computer user keeps audio files, either as MP3s, or in some other format on his/her computer's hard drive. It is very often that people upload/download music of various kinds, which requires a huge amount of bandwidth. This creates a need for better and better speech compression algorithms that reduces the size of the audio file significantly without sacrificing quality. Due to the increasing demand for better speech algorithms, several standards were developed, including MPEG, MP3, etc.

Data compression using transformations such as the DCT and the DFT are the basis for many coding standards such as JPEG, MP3 and AC-3. In this project FFT (IFFT) is used for the compression (decompression) of a speech signal. This data compression scheme is simulated using Matlab. Simulations are performed for different FFT sizes and different number of components chosen. Two different methods used for the purpose are:

- By retaining the first n-components
- By retaining dominant n-components

The SNR's (signal to noise ratios) are computed for all the simulations and used to study the behavior of the compression scheme using FFT. Also the noise introduced in the signal (for various cases) is studied both by listening to the recovered signal and by the calculated SNR's.

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## TECHNICAL BACKGROUND

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Fourier Transform (FT) can be very simply defined to be a mathematical technique to resolve a given signal into the sum of sines and cosines. The Fourier transform is an invaluable tool in science and engineering. The main features that make Fourier transform attractive are:

- Its symmetry and computational properties.
- Significance of time (space) vs. frequency (spectral) domain.

The Discrete Fourier Transform (DFT) is used to produce frequency analysis of discrete non-periodic signals. If we look at the equation for the Discrete Fourier Transform we will see that it is quite complicated to work out as it involves many additions and multiplications involving complex numbers. Even a simple eight-sample signal would require 49 complex multiplications and 56 complex additions to work out the DFT. At this level it is still manageable, however a realistic signal could have 1024 samples, which requires over 20,000,000 complex multiplications and additions. Obviously, this suggests that this technique becomes very time consuming with a slight increase in the number of samples.

The Fast Fourier Transform (FFT) is a discrete Fourier Transform (DFT) algorithm which reduces the number of computations from something on the order of  $N^2$  to  $N \log(N)$ . The Fast Fourier Transform greatly simplifies the computations for large values of  $N$ , where  $N$  is the number of samples in the sequence. The idea behind the FFT is the divide and conquer approach, to break up the original  $N$  point sample into two ( $N/2$ ) sequences. This is because a series of smaller problems is easier to solve than one large one. The DFT requires  $(N-1)^2$  complex multiplications and  $N(N-1)$  complex additions as opposed to the FFT's approach of breaking it down into a series of 2 point samples which only require 1 multiplication and 2 additions and the recombination of the points which is minimal. Two types of FFT algorithms are in use: decimation-in-time and decimation-in-frequency. The algorithm is simplified if  $N$  is chosen to be a power of 2, but it is not a requirement.

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## DETAILS OF THE PROGRAM

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Two main methods are implemented in the Matlab programs.

- By retaining the first  $n$ -components
- By retaining dominant  $n$ -components

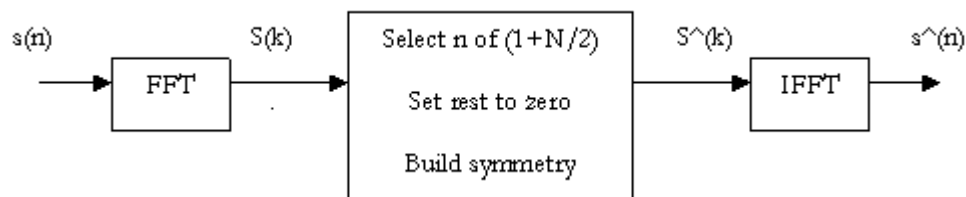
### RETAINING THE FIRST $N$ -COMPONENTS

In this method we start by reading the wave file “cleanspeech” in Matlab, and saving the speech in vector ‘s’. The data set (in the vector) to be compressed is then segmented into  $N$ -point segments (or frames) using a sliding window. This is done in the program by dividing the vector ‘s’ into segments. This is shown in the figure below.



*Figure 1: Sliding rectangular window*

The FFT of each segment is taken one by one by passing it into the loop  $N$  (the number of frames) times. Thus we get the magnitude spectrum of the signal. The result of the  $N$  point FFT has  $(1+N/2)$  independent components. This is due to the symmetry property of the DFT. These  $(1+N/2)$  points are retained as they are sufficient to get back the all the information in the original signal. From this set of about half the points first ‘ $n$ ’ points are chosen to reconstruct the original signal. In our simulation this is done for all possible  $n$  values from 1 to  $(1+N/2)$ . The rest of the  $(1+N/2-n)$  points are padded with zeros. Now, before taking the IFFT, we have to give the vector of ‘ $n$ ’ components its conjugate symmetry back. Otherwise, we will get back an imaginary signal. In order to rebuild the symmetry in the signal, the conjugate of the “first  $n$ -component” vector is taken. The first and last components in this new vector are disregarded as they are dc values which does not take part in the symmetry building before taking IFFT. The conjugate vector is flipped and added to the original “first  $n$ -component” vector. Hence, we have got the signal with symmetrical properties and we are ready to get back real values after taking the IFFT. The whole process is shown in the figure below.



*Figure 2: Data compression process*

The Matlab code for this method is provided in the appendix.

### RETAINING DOMINANT N-COMPONENTS

This method is very similar to the one we have discussed above. The only difference is in the way we select the 'n' points for the signal. In the previous case we chose the first 'n' components and set the rest to zero. In this case we will choose the dominant n points, i.e., the points with maximum magnitude. The rest of the  $(1+N/2-n)$  points are set to zero. Special care is taken to make the chosen dominant 'n' points lie at the indices they previously were (in the signal with  $1+N/2$  components). Also, in our program we have chosen our dominant signal to be at the minimum of the indices in case two components with the same indices are encountered. In this case the dominant point at the next index will be chosen in picking the following component (in case the n points are already not exhausted). The Matlab code for this method is provided in the appendix.

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**RESULTS**

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During the simulations we collected three sets of data, for 64, 128, and 256 point FFT. For each of the three sets 'n' (components selected) is varied from 1 to  $(1+N/2)$ . The tables summarizing these results follows.

*Table 1A: Simulation with N=64 (Rectangular window)*

n	N	Method 2	Method1
1	64	2.2038	0.0135
2	64	3.8719	0.0673
3	64	5.2410	0.1857
4	64	6.4056	0.4547
5	64	7.4867	1.4146
6	64	8.4827	3.9178
7	64	9.4121	5.3314
8	64	10.3473	6.1776
9	64	11.1988	6.6386
10	64	11.9796	7.1455
11	64	12.7710	7.9797
12	64	13.6107	9.0914
13	64	14.4537	11.2083
14	64	15.3109	11.9309
15	64	16.1816	12.2747
16	64	17.1226	12.5261
17	64	18.0293	12.7449
18	64	18.8967	13.0037
19	64	19.8869	13.4986
20	64	20.9640	14.6117
21	64	21.9970	16.2962
22	64	23.0627	17.6447
23	64	24.1988	18.9805
24	64	25.3514	20.0984
25	64	26.4813	21.1365
26	64	27.6716	22.1367
27	64	28.8479	23.2699
28	64	30.0733	24.5377
29	64	31.6307	26.0917
30	64	33.4667	28.7562
31	64	36.2471	32.2514
32	64	40.7178	37.5274
33	64	305.5805	305.5805

Table 2A: Simulation with  $N=128$  (Rectangular window)

n	N	Method 2	Method 1
1	128	2.3913	0.0032
2	128	3.7275	0.0108
3	128	4.7824	0.0370
4	128	5.6328	0.0556
5	128	6.3384	0.1102
6	128	6.9822	0.1977
7	128	7.5690	0.3129
8	128	8.1433	0.5590
9	128	8.6590	0.9857
10	128	9.1438	1.4231
11	128	9.6066	2.6956
12	128	10.0627	5.0373
13	128	10.5114	5.5290
14	128	10.9527	6.0540
15	128	11.3898	6.3331
16	128	11.8284	6.6376
17	128	12.2646	6.8445
18	128	12.7085	6.9992
19	128	13.1520	7.2733
20	128	13.5974	7.4969
21	128	14.0454	8.0543
22	128	14.4963	8.4776
23	128	14.9498	9.2602
24	128	15.3851	9.8516
25	128	15.7994	11.3802
26	128	16.2404	11.9881
27	128	16.6766	12.3644
28	128	17.0972	12.5158
29	128	17.5069	12.6674
30	128	17.9277	12.7595
31	128	18.3603	12.8306
32	128	18.8118	12.9309
33	128	19.2901	13.0164
34	128	19.7282	13.1321
35	128	20.2335	13.2215
36	128	20.7371	13.3550
37	128	21.2537	13.5309
38	128	21.7844	13.8250
39	128	22.3514	14.2770
40	128	22.8919	14.9846

41	128	23.4484	15.8839
42	128	23.9792	16.8314
43	128	24.5233	17.3377
44	128	25.0928	18.3420
45	128	25.6533	18.9407
46	128	26.2268	19.8718
47	128	26.8236	20.4231
48	128	27.4267	21.1078
49	128	28.0442	21.6144
50	128	28.6434	22.0653
51	128	29.2684	22.6249
52	128	29.9260	23.0933
53	128	30.6057	23.7241
54	128	31.2762	24.3549
55	128	31.9887	24.9883
56	128	32.7668	25.7346
57	128	33.6086	26.6174
58	128	34.5904	27.8736
59	128	35.7188	29.4067
60	128	36.8773	30.9742
61	128	38.5528	33.6740
62	128	40.2846	35.6910
63	128	43.1389	38.2483
64	128	47.9142	43.6936
65	128	305.6275	305.6275

Table 3A: Simulation with  $N=256$  (Rectangular window)

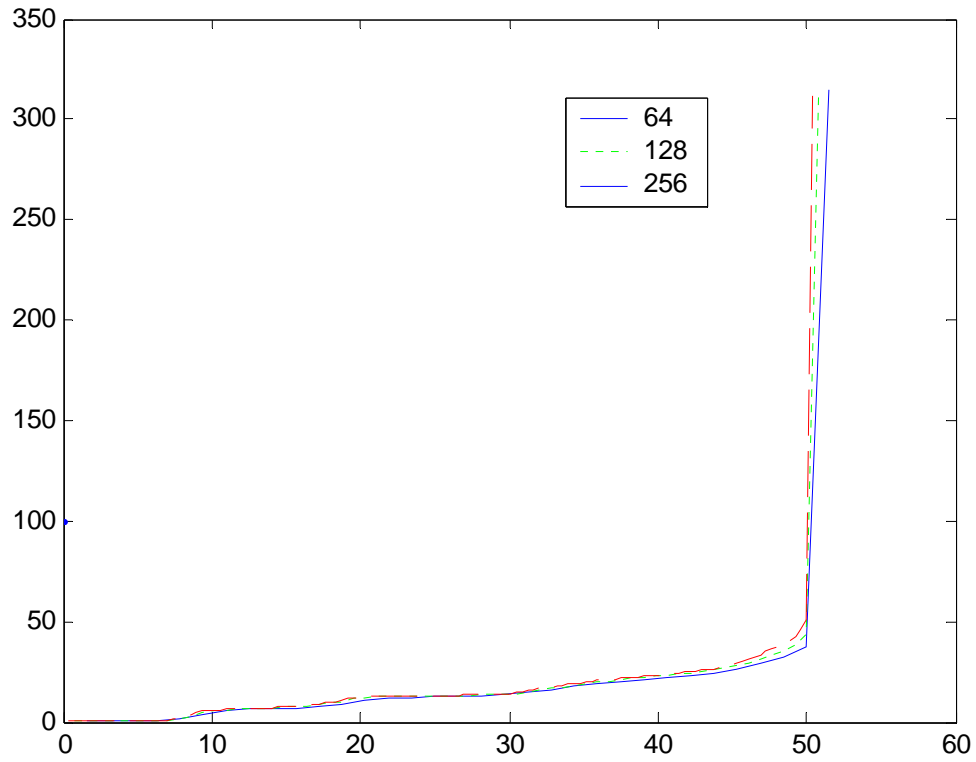
n	N	Method 2	Method1
1	256	2.0090	0.0007
2	256	3.0876	0.0023
3	256	3.7826	0.0041
4	256	4.3842	0.0096
5	256	4.9358	0.0247
6	256	5.4214	0.0280
7	256	5.8774	0.0323
8	256	6.3125	0.0590
9	256	6.6906	0.0859
10	256	7.0639	0.0987
11	256	7.4164	0.1197
12	256	7.7503	0.2151
13	256	8.0743	0.2684
14	256	8.3987	0.3077
15	256	8.7058	0.4391
16	256	9.0108	0.5822
17	256	9.3069	0.7697
18	256	9.5969	1.0126
19	256	9.8838	1.3183
20	256	10.1666	1.4826
21	256	10.4380	2.1441
22	256	10.6978	3.7475
23	256	10.9532	5.0288
24	256	11.1954	5.3707
25	256	11.4338	5.6791
26	256	11.6740	5.8293
27	256	11.9063	6.1157
28	256	12.1392	6.2348
29	256	12.3682	6.3834
30	256	12.5943	6.5169
31	256	12.8191	6.6682
32	256	13.0451	6.7548
33	256	13.2739	6.8624
34	256	13.4944	6.9606
35	256	13.7158	7.0365
36	256	13.9376	7.1189
37	256	14.1608	7.2260
38	256	14.3847	7.3879
39	256	14.6045	7.4739
40	256	14.8281	7.5760

41	256	15.0547	7.8927
42	256	15.2846	8.1763
43	256	15.5161	8.3581
44	256	15.7496	8.6720
45	256	15.9831	9.2201
46	256	16.2144	9.4267
47	256	16.4470	9.7869
48	256	16.6832	10.4543
49	256	16.9019	11.3228
50	256	17.1241	11.8626
51	256	17.3494	12.2138
52	256	17.5810	12.4037
53	256	17.8179	12.5534
54	256	18.0524	12.6431
55	256	18.2872	12.6970
56	256	18.5182	12.7878
57	256	18.7557	12.8467
58	256	18.9949	12.8774
59	256	19.2341	12.9083
60	256	19.4669	12.9584
61	256	19.7036	12.9878
62	256	19.9424	13.0258
63	256	20.1850	13.0690
64	256	20.4324	13.1144
65	256	20.6850	13.1563
66	256	20.9433	13.1994
67	256	21.2036	13.2470
68	256	21.4697	13.2825
69	256	21.7379	13.3318
70	256	22.0034	13.3786
71	256	22.2702	13.4342
72	256	22.5378	13.5005
73	256	22.8134	13.5983
74	256	23.0756	13.6950
75	256	23.3495	13.7923
76	256	23.6154	14.0204
77	256	23.8754	14.2565
78	256	24.1376	14.6320
79	256	24.4030	15.0195
80	256	24.6443	15.3790
81	256	24.9161	15.7558
82	256	25.1887	16.5182
83	256	25.4604	17.0393

84	256	25.7420	17.3047
85	256	26.0281	17.6261
86	256	26.3232	18.0332
87	256	26.6199	18.6680
88	256	26.9221	18.8803
89	256	27.2302	19.2169
90	256	27.5420	19.6178
91	256	27.8606	20.0707
92	256	28.1778	20.5241
93	256	28.5036	20.8405
94	256	28.8349	21.0330
95	256	29.1717	21.3219
96	256	29.5078	21.7062
97	256	29.8547	22.0363
98	256	30.2057	22.1975
99	256	30.5597	22.4879
100	256	30.9237	22.7819
101	256	31.2899	23.0213
102	256	31.6606	23.2471
103	256	32.0555	23.4681
104	256	32.4606	23.8325
105	256	32.8491	24.0880
106	256	33.2798	24.4810
107	256	33.7327	24.7880
108	256	34.1531	25.2040
109	256	34.5739	25.4550
110	256	35.0137	25.7486
111	256	35.4791	26.2008
112	256	35.9586	26.5935
113	256	36.4592	27.0083
114	256	36.9715	27.4299
115	256	37.4865	28.0088
116	256	38.0395	29.1043
117	256	38.6156	29.8402
118	256	39.2317	31.0556
119	256	39.8956	31.8653
120	256	40.5950	33.0205
121	256	41.3442	34.9470
122	256	42.2583	36.4348
123	256	43.3169	37.4570
124	256	44.5666	38.7509
125	256	46.1363	40.1556
126	256	48.1689	42.2606
127	256	50.8224	45.3212

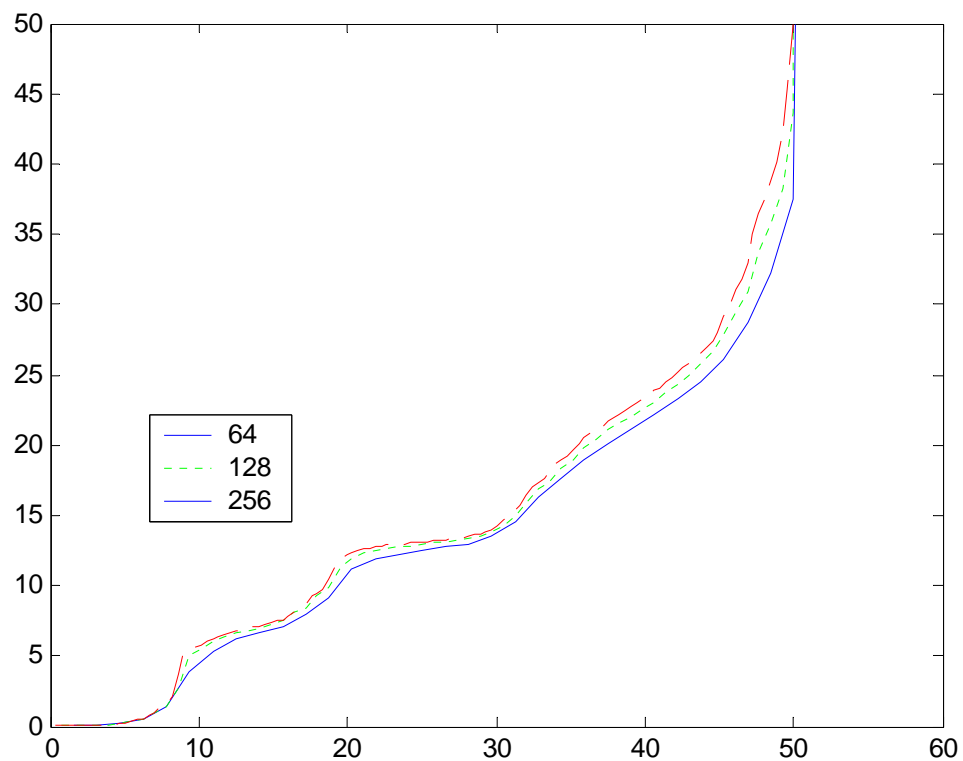
128	256	54.7673	50.2150
129	256	304.3307	304.3307

The data in the above tables was also plotted in three different ways. In the following figure the SNR curves for 64, 128, and 256 point FFT's are plotted on the same graph. The graph is for the method in which first 'n' components are selected is given in figure below.



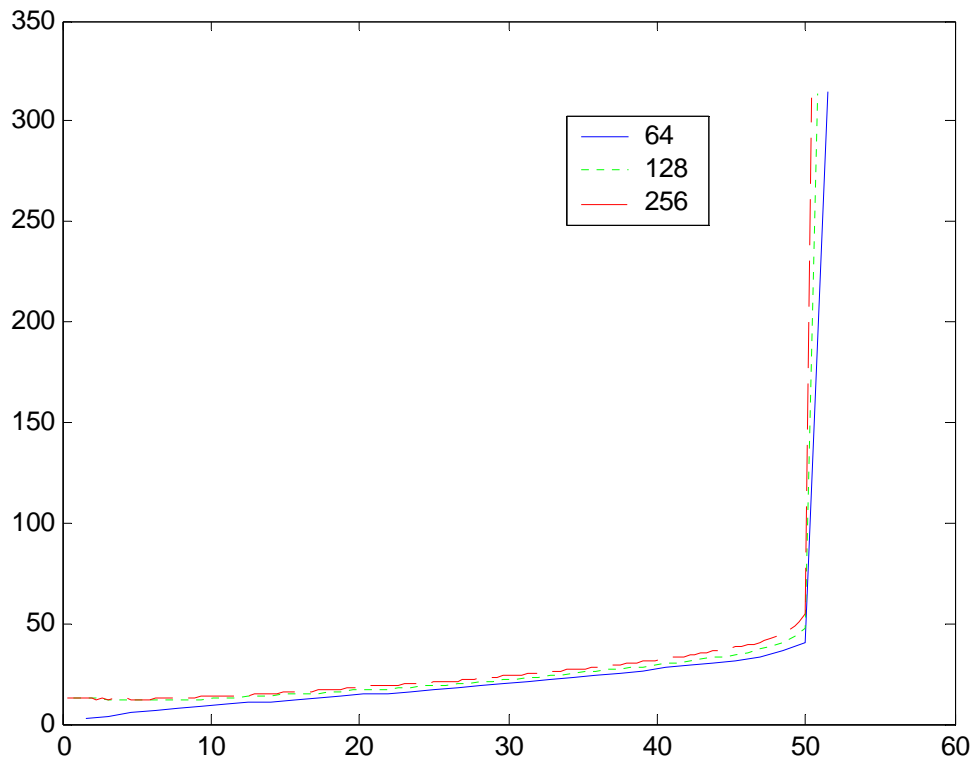
*Figure 3: SNR vs. percentage of components for first n-component method*

A second version of the same graph with scaled y-axis is given below.

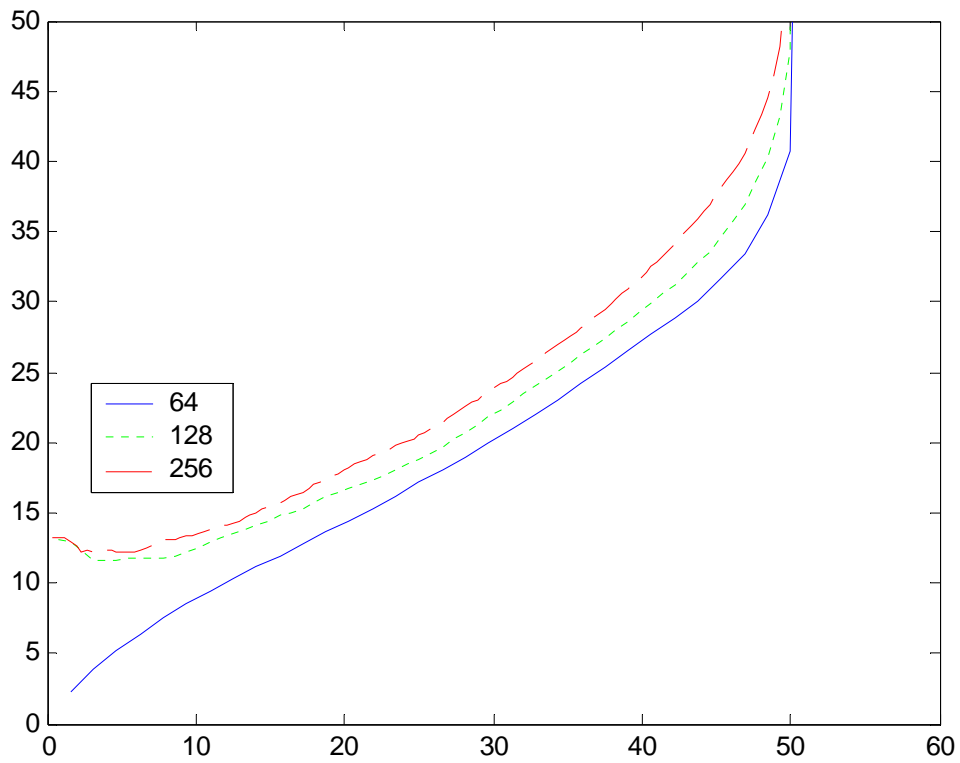


*Figure 4: SNR vs. percentage of components for first n-component method - Scaled Version*

In the following two figures the SNR curves for 64, 128, and 256 point FFT's are plotted on the same graph. The graphs are for the method in which dominant 'n' components are selected. The plot in second figure is a scaled version of the first to visualize the plot in a better way.

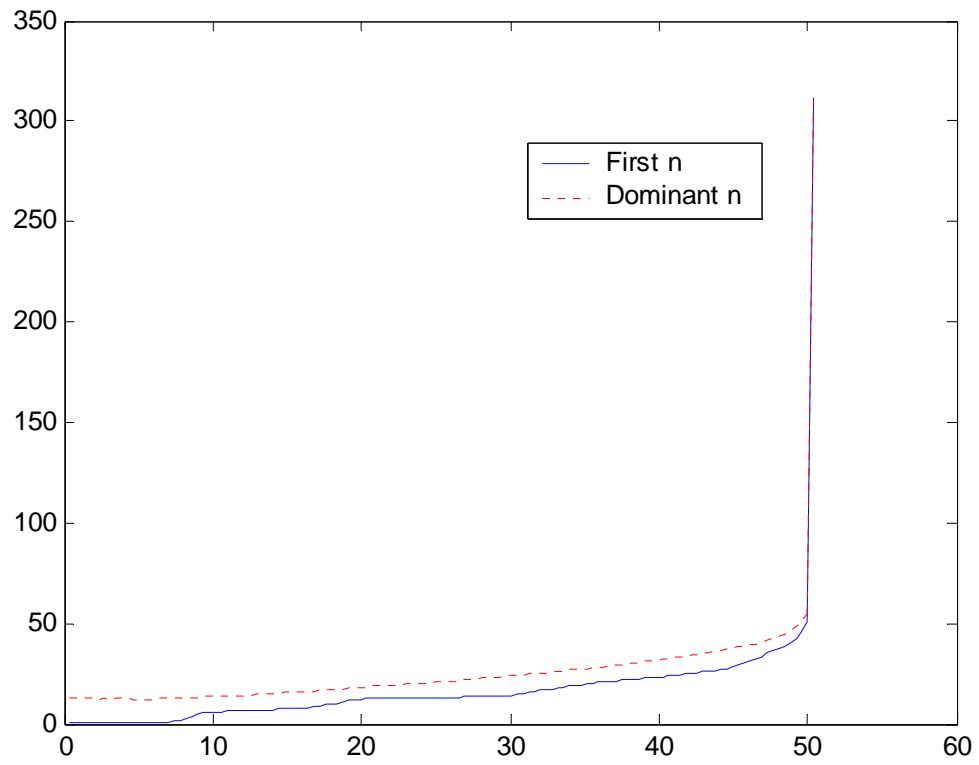


*Figure 5: SNR vs. percentage of components for dominant n-component method*

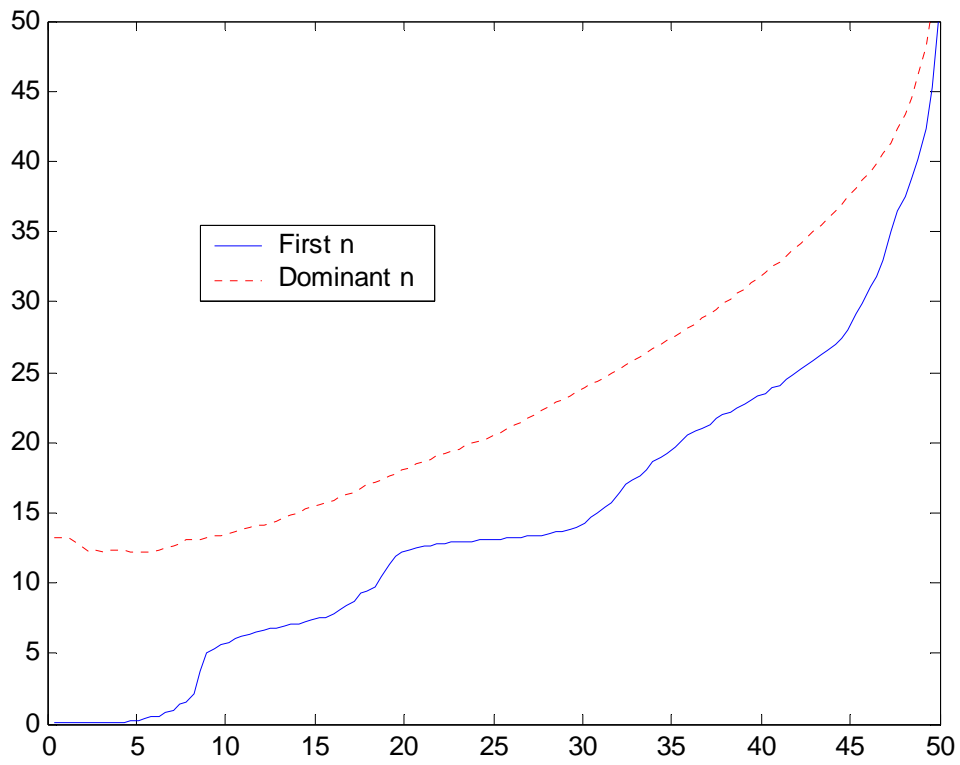


*Figure 6: SNR vs. percentage of components for dominant n-component method - Scaled Version*

In the following two plots SNR's (for the 256 point FFT) for first 'n' and dominant 'n' components are compared. The second plot is the scaled version of first.



*Figure 7: Comparison of the two methods for the case of  $N=256$*



*Figure 8: Comparison of the two methods for the case of  $N=256$  - Scaled Version*

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## REMARKS

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In this section we have answered the analysis questions.

1. What is the effect of the parameter  $n$  ( $N$ =fixed) on the SNR? Explain.

$N$  corresponds to the number of components of the signal chosen. This means that by increasing  $n$  value we are increasing the resolution of the signal, as we get closer to the original signal. This suggests that we should have an improvement in quality of sound as we increase  $n$ , both in the case of first  $n$  and dominant  $n$  methods. This is indeed the case and is supported by the audio signal created by the process. A signal with increased  $n$  gives a better quality audio signal. This can also be seen from the SNR values. The SNR values increase as we increase the  $n$  value from 1 to  $(1 + N/2)$ . The drawback of choosing large  $n$  is that the size of the file starts getting bigger as we increase  $n$ .

2. What is the effect of  $N$  ( $n/N$ =fixed) on the SNR? Explain.

$N$  in our program refers to the size of FFT used. This is in fact also the length of the window used for the simulation of a particular  $N$  sized FFT. We have done simulations for three values of  $N$ , namely 64, 128, and 256. As we increase the value  $N$ , we increase the resolution of our signal, by increasing the number of samples. This will increase the quality of the audio signal. In our case the quality of the audio signal simultaneously depends on  $N$  and  $n$  values. So if  $N$  value is increased but  $n$  value is chosen to be very low, the overall signal will not be a high quality signal. Choosing  $N$  to be 64, our best quality compressed signal will be composed of 33 non-zero components. From best quality we mean that  $n$  is chosen to be at its peak value. In the case of  $N=128$ , our best quality signal will be composed of 65 non-zero components. In the case of  $N$  being 256, the best quality compressed signal will have 129 non-zero components.

3. Explain the differences in the results obtained with method 1 as opposed to method 2.

Using first  $n$  component method usually provides a relatively poor result compared to the results provided by the method of choosing  $n$ -dominant components. This can be seen from the SNR plots. This should also be our intuitive answer, as by choosing the  $n$  dominant points we are in fact taking account of a wider range of values. Since these values are picked so that they have high magnitudes, the quality of audio is relatively better. Choosing the first  $n$  components might provide us with non-useful information cutting out the important part of the signal. Using the dominant  $n$ -component method we have a smaller chance of getting into such situations. We also note that the SNR values turn out to be the same for a particular  $N$  and maximum possible  $n$ . This should indeed be the case as when we choose maximum possible  $n$ , the components from  $n$  dominant and first  $n$  methods should be identical.

4. In order to implement an actual data compression scheme then the retained transform components must be encoded in binary format. Assuming that  $n$  and  $N$  are the same for method 1 and method 2, which method will produce the lowest bit-rate (bits/second)?

The lowest bit-rate will be provided by the method in which we choose the first  $n$ -components. This is because we have higher magnitudes in the case of dominant  $n$ -component method. On average each component of “ $n$  dominant component” method will have greater magnitude than the component of “first  $n$ -component” method. This suggests that a greater bit-rate is needed for dominant  $n$ -component method. In changing the signal components to binary format we will have lower bit rates for “first  $n$ -component” method. For example, we can represent ‘1’ as ‘01’ in binary, but to represent 8 we have to have at least 3 bits, that is, ‘111’.

5. Try to listen to the processed files using the MATLAB sound command and give some comments regarding the subjective quality of the processed record.

For low values of  $n$ , keeping the  $N$  constant, the quality of voice obtained with  $n$  dominant component method is much better. The actual voice (information bearing) part of the signal is clearer in this case. In the case of first  $n$ -component method it is difficult to distinguish between noise and voice. It seemed that the voice signal obtained by the first  $n$ , and dominant  $n$  components can be compared to AM and FM radio respectively. When choosing  $n$  to be in the mid-range values, the “first  $n$ -component” method produced a voice signal which has more noise than the other case, but it sounded more smooth than the other case. This is because we found sort of “clipping” in the signal produced by choosing dominant components. At high values of  $n$  the voice signals generated by using the two different methods sounded almost identical.

This has been a wonderful learning experience. Specially listening to the compressed file and figuring out what effects does the two methods on the quality of sound was particularly interesting. We learnt how to do some serious work in Matlab. We learnt and were amazed by the possibilities Matlab programming provides us with (in order to do mathematical operations).

I think that providing students with such examples of code and letting them play around with the code can be useful for the students. A lab can be made where this or a similar kind of code example can be given to the students. It will be useful and fun to answer questions similar to the ones asked in this project. I think that it can be a very good learning experience because it is not something that is purely mathematical, but the students will in fact be able to experience a worldly example or application of DSP.

Same amount of time and effort has been put into this project by the two team members. We both came up with a code for first  $n$ -component method separately. The only problem was that one of us was not getting the correct result at the value when  $n=1+N/2$ . We figured the dominant  $n$ -component method by sitting together and discussing what can be changed in the first code to make it work for the dominant case. Introduction, technical background, and the data tables are written and prepared by Hassan Mansoor. The details of the program, plots, and remarks section is prepared by Imtiaz Nizami.

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## APPENDIX

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### Code to implement the method with first n-components:

```
clear,clc;
hold off
tic
for fftPoints= 1 : 3
    switch fftPoints
    case 1
        N=64;
        M=64;
    case 2
        N=128;
        M=128;
    case 3
        N=256;
        M=256;
    end

    s=wavread('cleanspeech');           % Load wave file into matlab as vector
    L=length(s);                       % Scalar representing length of wave-file vector
    S=zeros(N,1);

    S1=zeros(1+N/2,1);
    S2=zeros(1+N/2,1);
    S3=zeros(1,N);
```

```

S4=zeros(1,N);

S5=zeros(L,1);

for i0 = 1:1+N/2

    N1=1:i0;

    for i = 1:K

        S1=zeros(1+N/2,1);

        k=(1:M)+((i-1)*M);

        k=min(k):(min(max(k),L));

        S=fft(s(k),N);

        S1(N1)=S(N1);

        S2=flipud(conj(S1));

        S3=[S1;S2(2:N/2)];

        S4=ifft(S3,N);

        S5(k)=S4;

    end

    S6=real(S5);

    index=1:L;

    S11=s(index);

    S12=S6(index);

    sum(S11.^2);

    sum(S11-S12).^2;

    SNR(i0)=10*log10(sum(S11.^2)./sum((S11-S12).^2));

end

```

```

switch fftPoints
case 1
    SNR_64_1=SNR;
case 2
    SNR_128_1=SNR;
case 3
    SNR_256_1=SNR;
end
end

n1=1:33;
n2=1:65;
n3=1:129;
plot(n1*100/64,SNR_64_1)
hold on;
plot(n2*100/128,SNR_128_1,'g--')
plot(n3*100/256,SNR_256_1,'r-')
toc

```

**Code to implement the method with dominant n-components:**

```

clear,clc;
hold off
tic

for fftPoints= 1 : 3

```

```

switch fftPoints
case 1
    N=64;
    M=64;
case 2
    N=128;
    M=128;
case 3
    N=256;
    M=256;
end

s=wavread('cleanspeech');           % Load wave file into matlab as vector
L=length(s);                       % Scalar representing length of wave-file vector
K=round(L/M);                       % Total number of frames
S_old=zeros(N,1);
S=zeros(1+N/2,1);
abs_S=zeros(1+N/2,1);

S1=zeros(1+N/2,1);
S2=zeros(1+N/2,1);
S3=zeros(1,N);
S4=zeros(1,N);
S5=zeros(L,1);

for i0 = 1:1+N/2

```

```

for i = 1:K
    k=(1:M)+((i-1)*M);
    k=min(k):(min(max(k),L));
    S_old=fft(s(k),N);
    S=S_old(1:1+N/2);
    abs_S=abs(S);
    min_abs_S=0;

    S1=zeros(1+N/2,1);

    for count1 = 1:i0
        max_index=find(abs_S==max(abs_S));
        index_value(count1)=min(max_index);
        abs_S(min(max_index))=min_abs_S;
    end

    S1(index_value)=S(index_value);
    S2=flipud(conj(S1));
    S3=[S1;S2(2:N/2)];
    S4=ifft(S3,N);
    S5(k)=S4;
end

S6=real(S5);
index=1:L;
S11=s(index);

```

```

    S12=S6(index);

    sum(S11.^2);

    sum((S11-S12).^2);

    SNR(i0)=10*log10(sum(S11.^2)./sum((S11-S12).^2));

end

switch fftPoints

case 1

    SNR_64_2=SNR;

case 2

    SNR_128_2=SNR;

case 3

    SNR_256_2=SNR;

end

end

n1=1:33;

n2=1:65;

n3=1:129;

plot(n1*100/64,SNR_64_2)

hold on;

plot(n2*100/128,SNR_128_2,'g--')

plot(n3*100/256,SNR_256_2,'r-')

toc

```