

Complexity Analysis of Discrete Fourier Transform Algorithms

Final Project Presentation
Data Structures and Algorithms (ECE 573)

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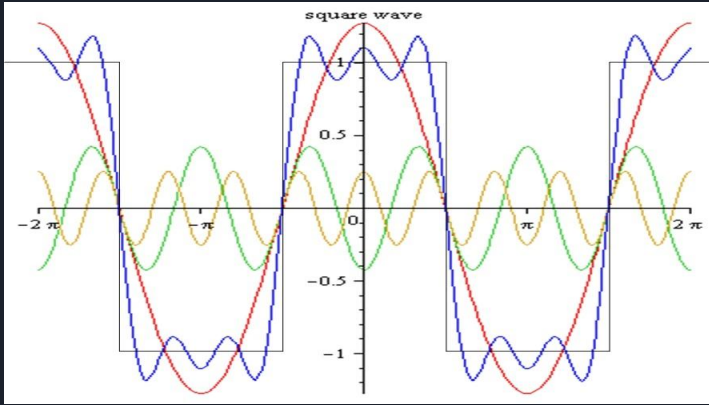


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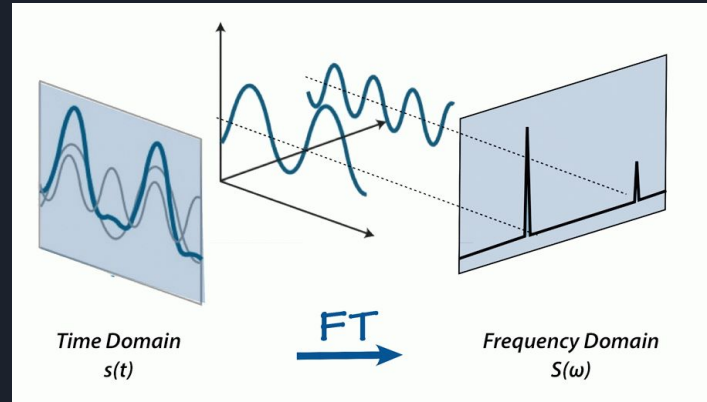
Introduction

- Fourier Transform developed by French mathematician Jean-Baptiste Joseph Fourier in late 1700's



- Any continuous function or signal can be decomposed as sum of infinite sines and cosines or complex exponentials

- Fourier Transform quantifies / measures the frequency content in a signal





Introduction

- Applications of Fourier Transform:
 - Spectrum Visualization
 - Simplifying complex operations like convolutions
 - Design and Implementation of Filters
 - Modern Communication Systems (OFDM)
- Discrete Fourier Transform was developed as the digital counterpart to the original Fourier Transform
- The Discrete Fourier Transform not feasible for real time implementations
 - Complexity $O(N^2)$
- The Fast Fourier Transform is a smarter implementation of the DFT
 - Complexity of $O(N \log(N))$ 100x faster than DFT



Algorithms: Discrete Fourier Transform

- The Discrete Fourier Transform is defined as follows:

$$X_k = \sum_{n=0}^{N-1} x_n e^{-\frac{2\pi i}{N} nk}$$

- Input $\rightarrow x(n)$, DFT $\rightarrow X(k)$ both length N
- N^2 number of complex multiplications and $N(N-1)$ complex additions ($O(N^2)$ complexity)

Algorithms: Radix-2 Fast Fourier Transform

$$X_k = \sum_{n=0}^{N-1} x_n e^{-\frac{2\pi i}{N} nk}$$

- Input $\rightarrow x(n)$, FFT $\rightarrow X(k)$ both length N

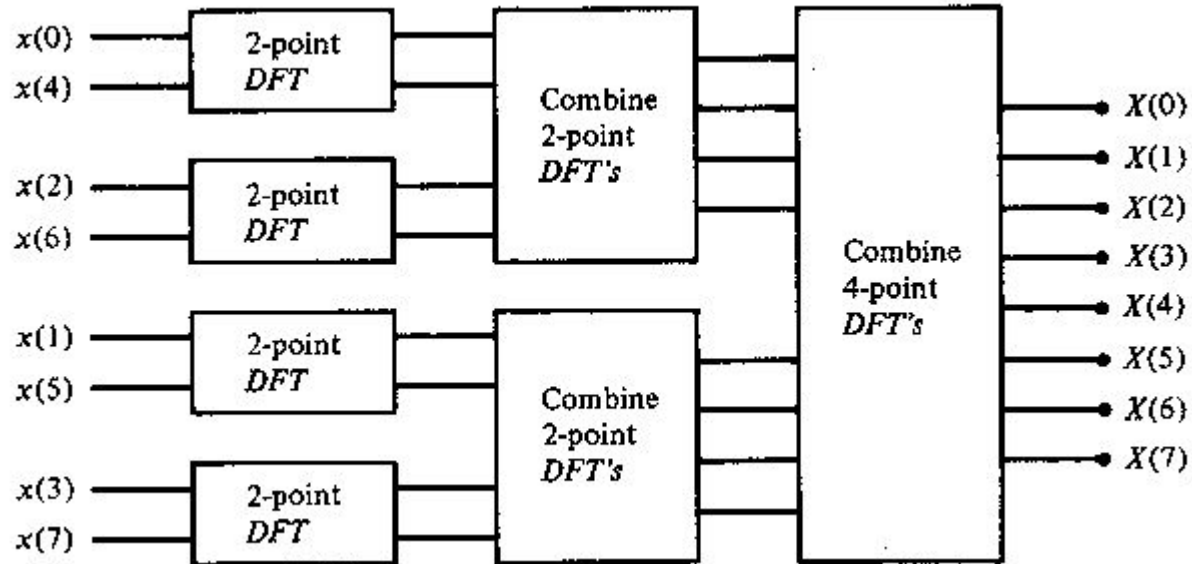
$$X_k = \sum_{m=0}^{\frac{N}{2}-1} x(2m) \cdot W_N^{m,k} + W_N^k \cdot \sum_{m=0}^{\frac{N}{2}-1} x(2m+1) \cdot W_N^{m,k}$$

- Split input into even, odd halves and multiply twiddle factor with odd seq

$$X_k = X_{k(\text{even})} + W_N^k \cdot X_{k(\text{odd})}, \quad k = 0 \dots \frac{N}{2} - 1$$

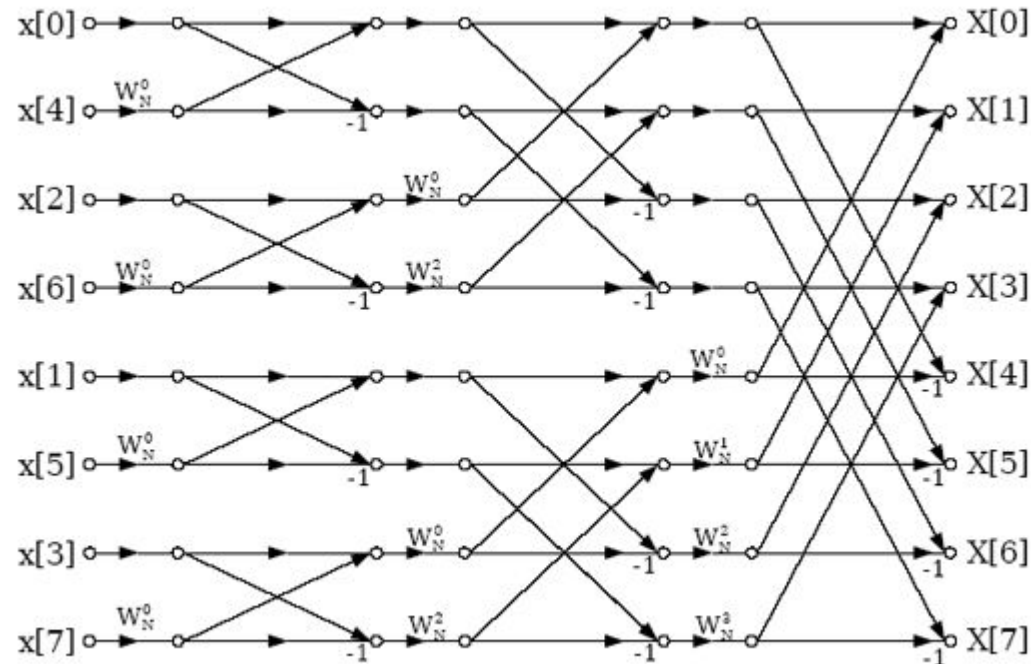
- $(N/2)^2$ number of complex multiplications reduces to $O(N \log(N))$ complexity

Algorithms: Radix-2 Fast Fourier Transform



Decimation in time

Algorithms: Radix-2 Fast Fourier Transform



Decimation in time



Experiment

Procedure

- DFT and FFT implemented using Python
- 4 Datasets generated for input
 - Zeros
 - Ones
 - -1 to 1
 - Cosine
- Problem sizes range from 256 (2^8) up to 32K (2^{15})
- Performance evaluated by 2 criteria
 - Execution time
 - Number of operations

Hardware

- Experiments run on 2 machines
 - 64-bit Ubuntu platform running Linux OS
 - 64-bit Windows platform running virtual machine of Linux OS

Simulations

- 128 simulations performed
 - 2 Algorithms
 - 4 Datasets
 - 8 Problem Sizes
 - 2 Machines



Results

Dataset: Ones

Problem Size		DFT (Ubuntu VM)	DFT (Ubuntu)	FFT (Ubuntu VM)	FFT (Ubuntu)
256	2^8	0.11174	0.04728	0.00218	0.0063
512	2^9	0.29339	0.181	0.00849	0.001
1k	2^10	1.12745	0.72603	0.02346	0.016
2k	2^11	3.75467	2.85524	0.04848	0.0234
4k	2^12	14.83544	11.36227	0.09605	0.0379
8k	2^13	97.58259	52.01844	0.17482	0.0708
16k	2^14	359.97122	230.09249	0.2375	0.1477
32k	2^15	1160.80973	855.50801	0.4751	0.3079

Problem Size		DFT (Ubuntu VM)	DFT (Ubuntu)	FFT (Ubuntu VM)	FFT (Ubuntu)
256	2^8	0.12363	0.05373	0.0025	0.0067
512	2^9	0.29977	0.18792	0.0086	0.0125
1k	2^10	1.0552	0.70518	0.0204	0.013
2k	2^11	4.10548	2.81096	0.0503	0.0207
4k	2^12	15.2145	11.1803	0.1013	0.0371
8k	2^13	89.88514	50.29091	0.1536	0.0703
16k	2^14	395.60079	224.86547	0.33671	0.14658
32k	2^15	1265.86637	846.19161	0.7582	0.309

Dataset: Zeros



Results

Dataset: -1 to 1

Problem Size		DFT (Ubuntu VM)	DFT (Ubuntu)	FFT (Ubuntu VM)	FFT (Ubuntu)
256	2^8	0.1307	0.05571	0.00241	0.0054
512	2^9	0.29633	0.17978	0.00841	0.009
1k	2^10	0.96175	0.70308	0.026	0.017
2k	2^11	5.03086	2.77729	0.0589	0.0203
4k	2^12	22.94392	11.08887	0.09496	0.0407
8k	2^13	88.62576	52.68572	0.1955	0.0719
16k	2^14	356.47654	221.95255	0.2547	0.1522
32k	2^15	1204.3287	879.47403	0.4598	0.3054

Problem Size		DFT (Ubuntu VM)	DFT (Ubuntu)	FFT (Ubuntu VM)	FFT (Ubuntu)
256	2^8	0.10748	0.05634	0.0022	0.0062
512	2^9	0.28475	0.18265	0.0053	0.0038
1k	2^10	1.06186	0.70221	0.00949	0.0158
2k	2^11	4.9707	2.8	0.03841	0.0226
4k	2^12	16.20032	11.1536	0.09208	0.0397
8k	2^13	82.95861	50.31715	0.15306	0.0734
16k	2^14	353.26238	225.3965	0.2559	0.1484
32k	2^15	0.10748	0.05634	0.0022	0.0062

Dataset: Cosine



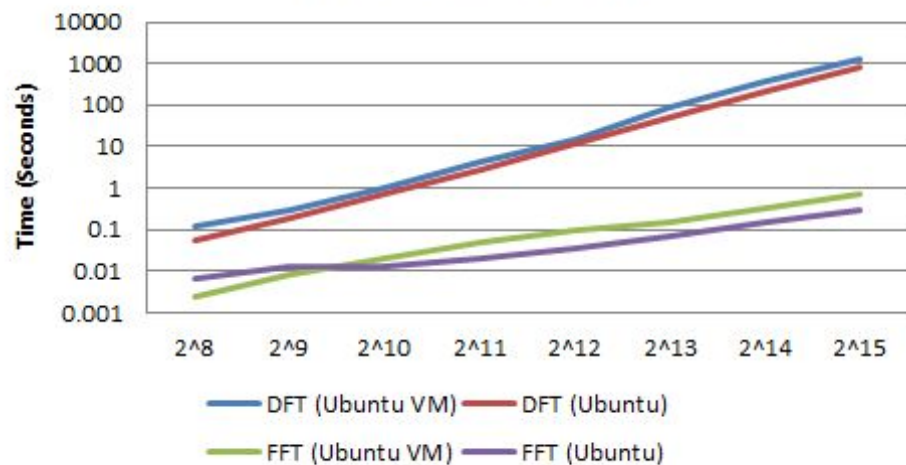
Results

Number of Operations Performed

Problem Size		DFT	FFT
256	2^8	65536	2048
512	2^9	262144	4608
1k	2^{10}	1048576	10240
2k	2^{11}	4194304	22528
4k	2^{12}	16777216	49152
8k	2^{13}	67108864	106496
16k	2^{14}	268435456	229376
32k	2^{15}	1073741824	491520

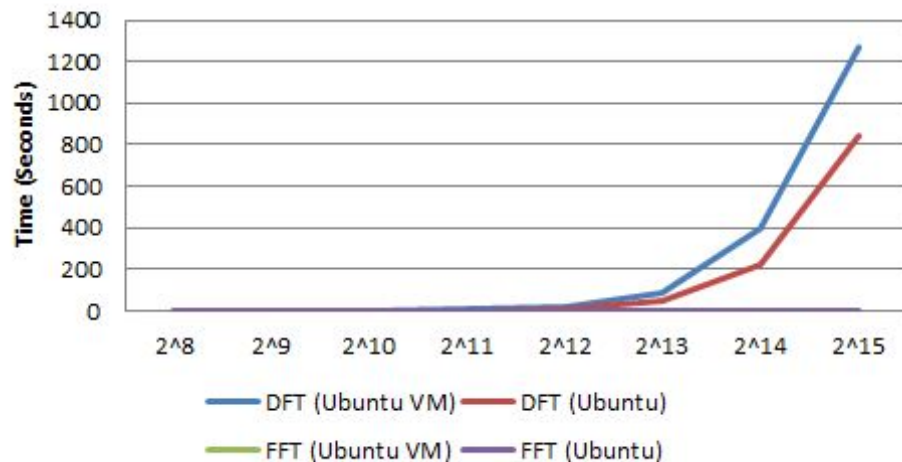
Results

Execution Time: Zeros



Logarithmic Scale

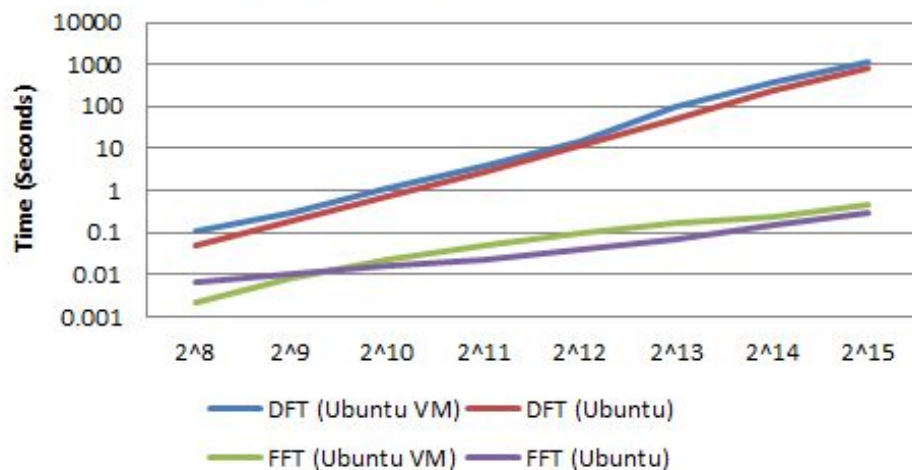
Execution Time: Zeros



Linear Scale

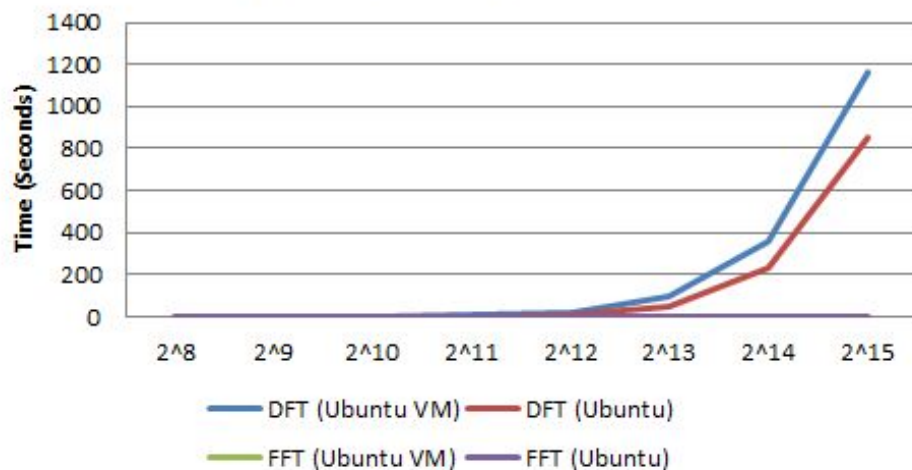
Results

Execution Time: Ones



Logarithmic Scale

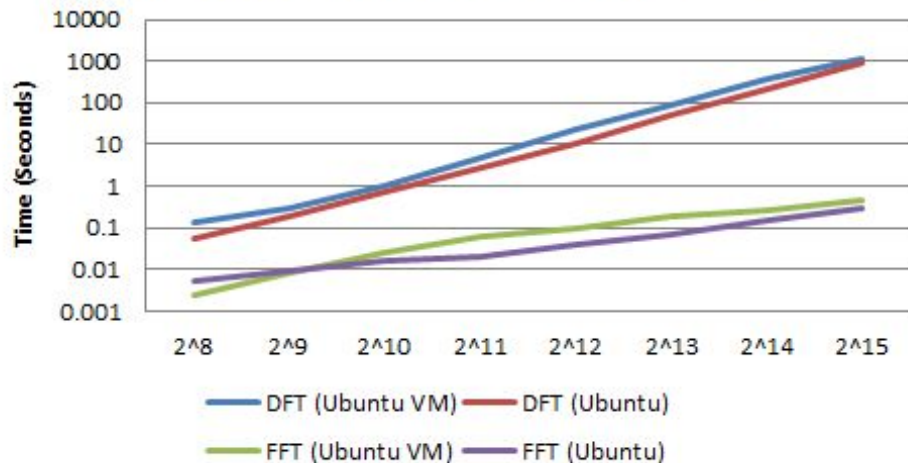
Execution Time: Ones



Linear Scale

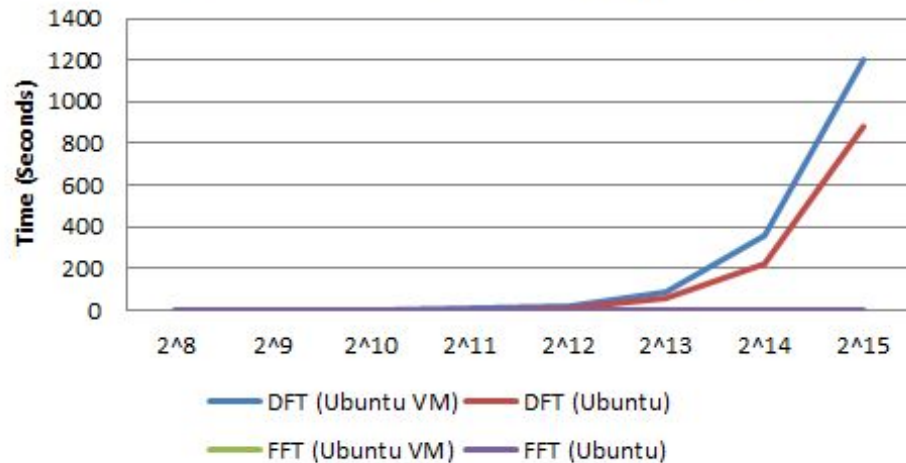
Results

Execution Time: Alternating -1 to 1



Logarithmic Scale

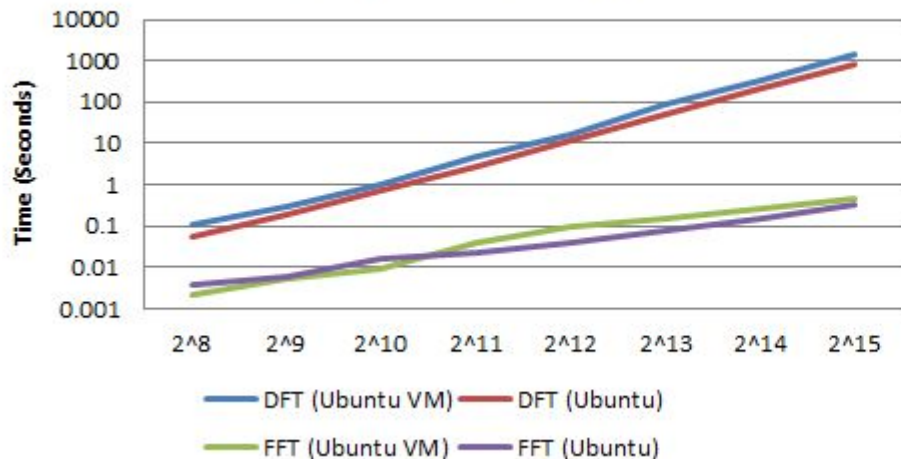
Execution Time: Alternating -1 to 1



Linear Scale

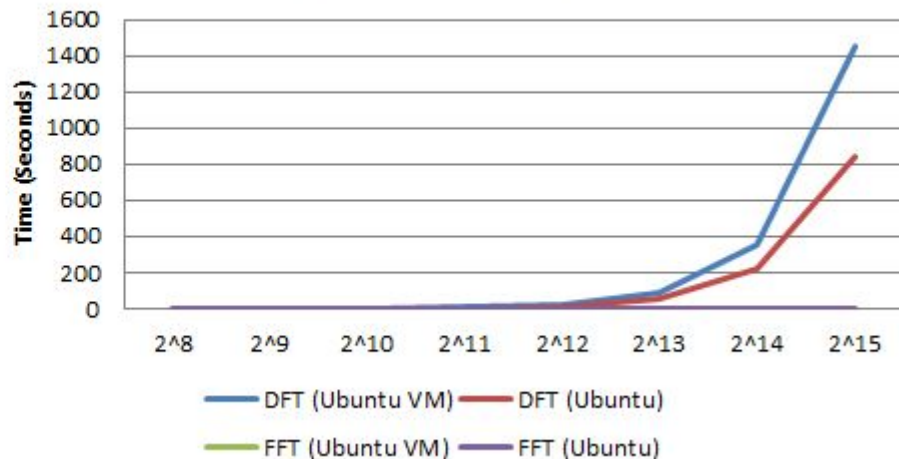
Results

Execution Time: Cosine



Logarithmic Scale

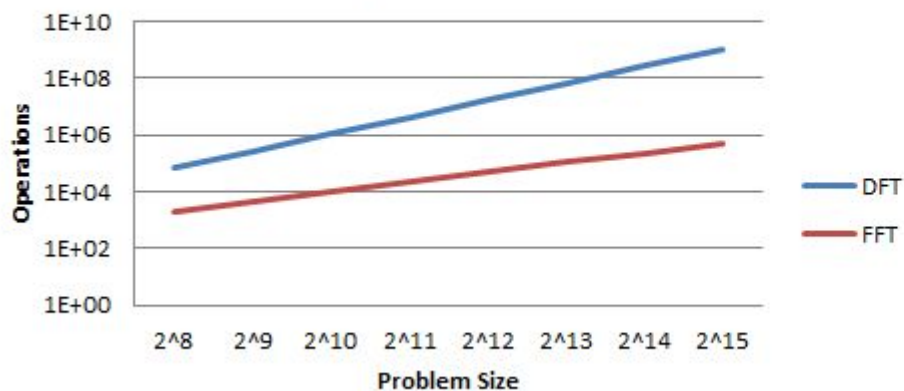
Execution Time: Cosine



Linear Scale

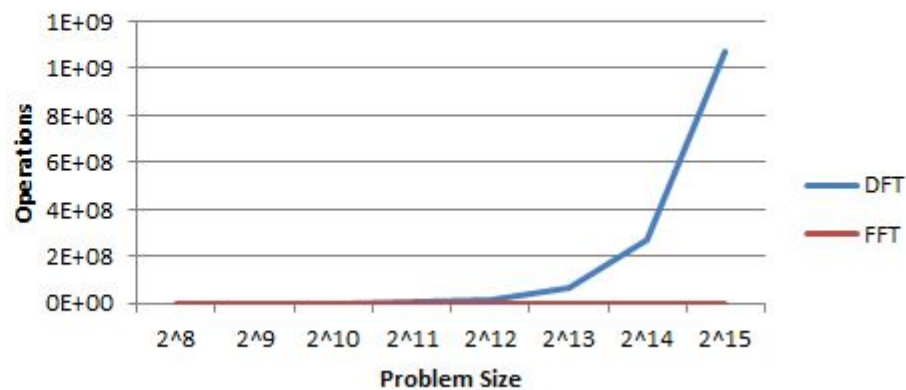
Results

Problem Size vs. Number of Operations



Logarithmic Scale

Problem Size vs. Number of Operations



Linear Scale



Conclusions

- 10x to <1000x reduction in number of operations
- 10x to >1000x increase in speed
- DFT (of length N) calculated by a summation of input sequence (length N) elementwise
 - $\rightarrow O(N^2)$ complexity
- Radix-2 FFT recursively halves input sequence (length N) to compute DFT on halved sequence
 - $\rightarrow O(N \log(N))$ complexity

QUESTIONS?

