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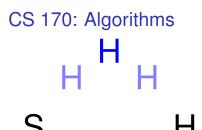


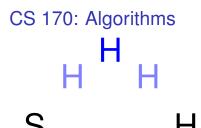






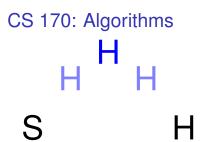
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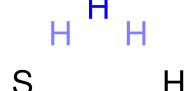


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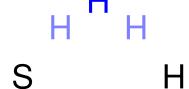
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No laptops please.



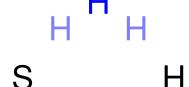
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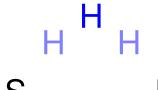
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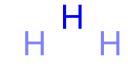
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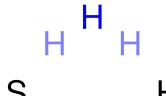
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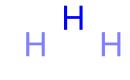
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Thank you!!!!



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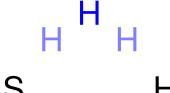
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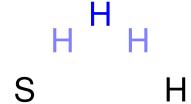
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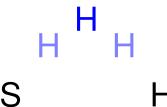
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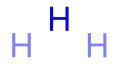
Thank you!!!!!!!!



No laptops please.

Thank you!!!!!!!!

Today:



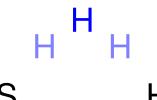
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Thank you!!!!!!!!

Today:

Review Inverse FFT



the second secon

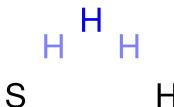
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Thank you!!!!!!!!

Today:

Review Inverse FFT

FFT network.



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Thank you!!!!!!!!

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Review Inverse FFT

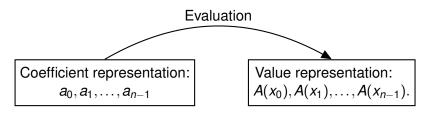
FFT network.

Other parallel systems.

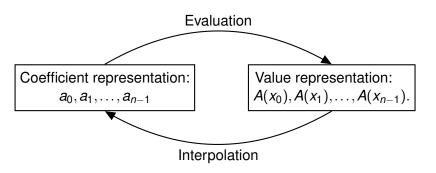
Coefficient representation: $a_0, a_1, ..., a_{n-1}$

Value representation: $A(x_0), A(x_1), \dots, A(x_{n-1}).$

Evaluation: $O(n \log n)$ if choose $1, \omega, \omega^2, \dots, \omega^{n-1}$.

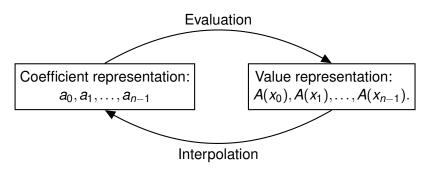


Evaluation: $O(n \log n)$ if choose $1, \omega, \omega^2, \dots, \omega^{n-1}$.



Interpolation: From points $A(x_0),...,A(x_{n-1})$ to "function".

Evaluation: $O(n \log n)$ if choose $1, \omega, \omega^2, \dots, \omega^{n-1}$.



Interpolation: From points $A(x_0), \dots, A(x_{n-1})$ to "function". How?

Compute $A(\cdot)$ from a_i 's:

$$\begin{bmatrix} A(x_0) \\ A(x_1) \\ \vdots \\ A(x_{n-1}) \end{bmatrix} = \begin{bmatrix} 1 & x_0 & x_0^2 & \cdots & x_0^{n-1} \\ 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ & \vdots & \ddots & \ddots \\ 1 & x_{n-1} & x_{n-1}^2 & \cdots & x_{n-1}^{n-1} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix}$$

Compute inverse of matrix above.

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Compute inverse of matrix above. Multiply.

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Doh!!

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Doh!!

Also, computing inverse not even easy.

FFT Matrix

FFT: ω is complex \emph{n} th root of unity

FFT Matrix

FFT: ω is complex *n*th root of unity and matrix is ...

T:
$$\omega$$
 is complex n th root of unity dimatrix is ...
$$M_n(\omega) = \begin{bmatrix} 1 & 1 & 1 & \cdots & 1 \\ 1 & \omega & \omega^2 & \cdots & \omega^{(n-1)} \end{bmatrix}$$

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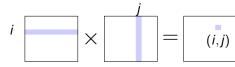
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Compute inverse of $M_n(\omega)$?

Inversion formula: $(M_n(\omega))^{-1} = \frac{1}{n} M_n(\omega^{-1})$.

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FFT works with points with basic root of unity: ω or ω^{-1}

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Evaluation: $FFT(A, \omega)$.

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FFT works with points with basic root of unity: \omega or \omega^{-1} 1, \omega^{-1}, \omega^{-2}, \dots, \omega^{-(n-1)}. \omega^{-1} is a primitive nth root of unity!
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Fast convolution! Digital signal processing.

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Fast convolution! Digital signal processing. Many other things.

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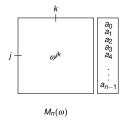
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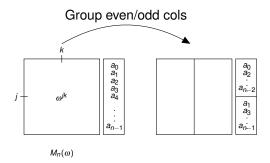
Interpolation: $\frac{1}{n}$ FFT(A, ω^{-1}).

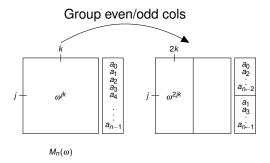
 \implies $O(n \log n)$ time for multiplying degree n polynomials.

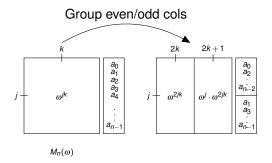
Fast convolution! Digital signal processing. Many other things.

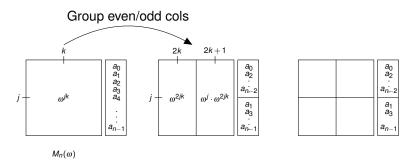
Cool!!

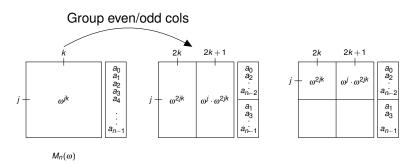


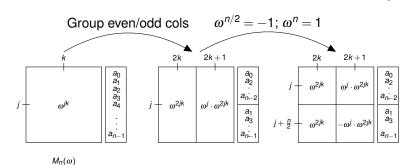


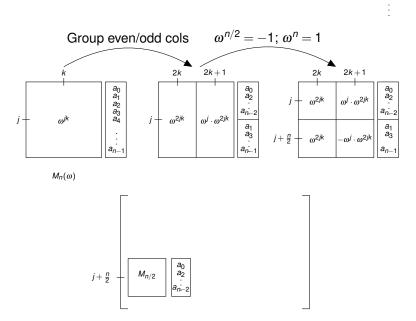


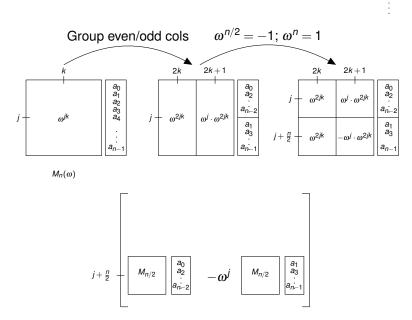


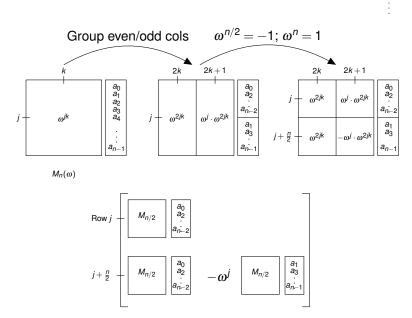


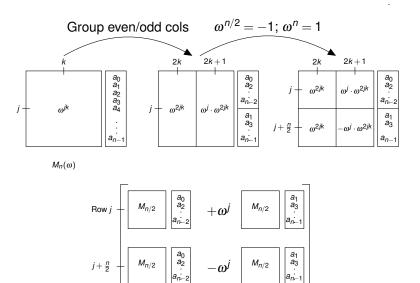


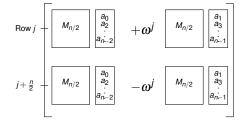




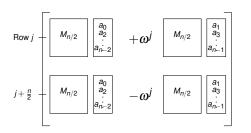


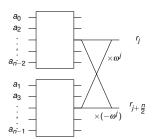


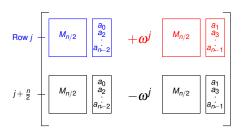


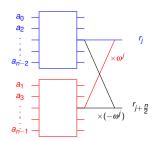


Butterfly switches!

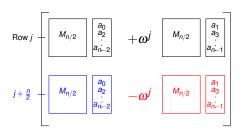


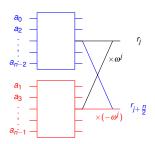




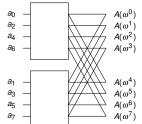


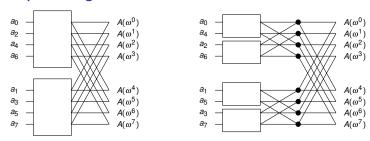
Butterfly switches!

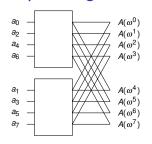


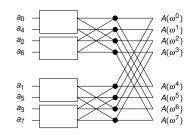


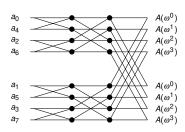
Butterfly switches!

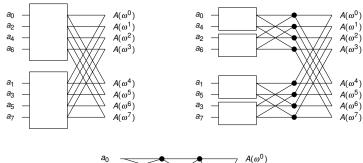


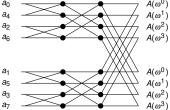




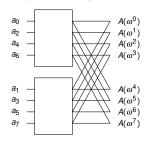


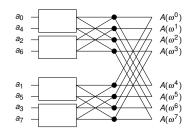


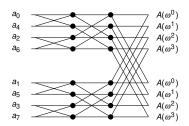




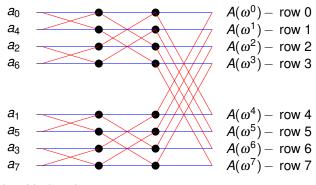
Edges from lower half of FFT have multipliers!



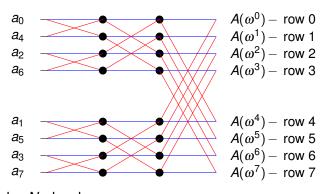




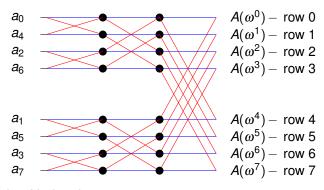
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log N - levels.



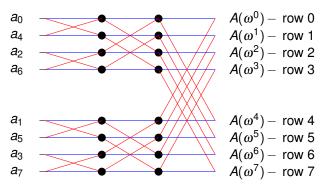
log N - levels. N - rows.



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In level i:

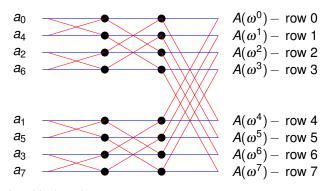


log N - levels.

N - rows.

In level i:

Row r node is connected to row r node in level i + 1.



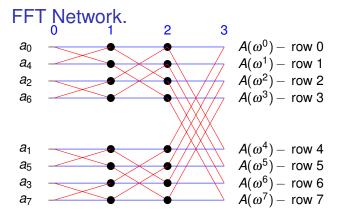
log N - levels.

N - rows.

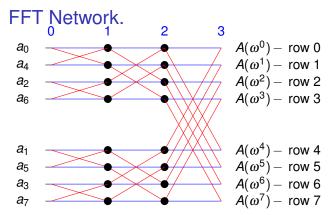
In level i:

Row r node is connected to row r node in level i + 1.

Row *r* node connected to row $r \pm 2^i$ node in level i + 1

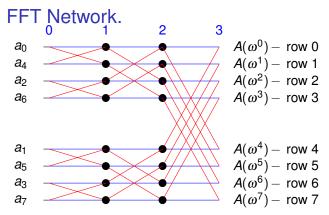


Row *r* node connected to row $r \pm 2^i$ node in level i + 1



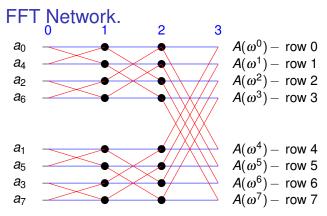
Row r node connected to row $r \pm 2^i$ node in level i + 1When is it $r + 2^i$?

- (A) When $|r/2^i|$ is odd.
- (B) When $\lfloor r/2^i \rfloor$ is even.



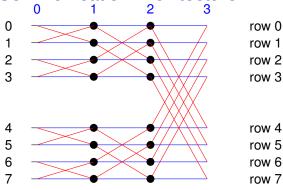
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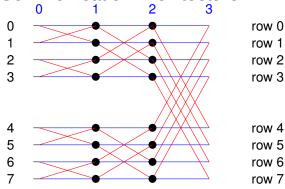
- (A) When $|r/2^i|$ is odd.
- (B) When $|r/2^i|$ is even.
- (B).



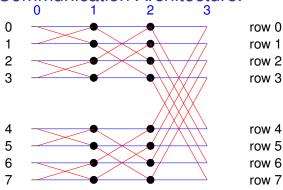
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- (A) When $|r/2^i|$ is odd.
- (B) When $|r/2^i|$ is even.
- (B). Red edges flip bit!



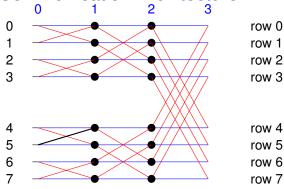


Route from input i = 101 to output j = 000?



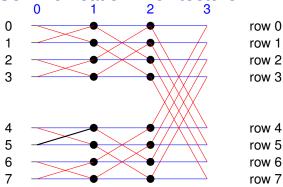
Route from input i = 101 to output j = 000?

Flip first bit.



Route from input i = 101 to output j = 000?

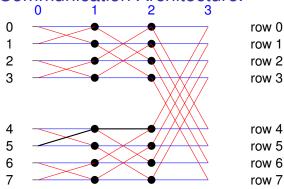
Flip first bit. Red (cross) edge.



Route from input i = 101 to output j = 000?

Flip first bit. Red (cross) edge.

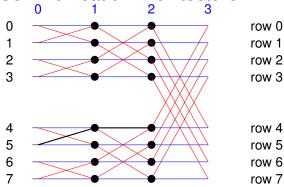
Keep second bit.



Route from input i = 101 to output j = 000?

Flip first bit. Red (cross) edge.

Keep second bit. Blue (straight) edge.

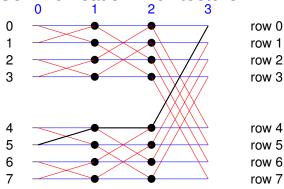


Route from input i = 101 to output j = 000?

Flip first bit. Red (cross) edge.

Keep second bit. Blue (straight) edge.

Flip third bit.

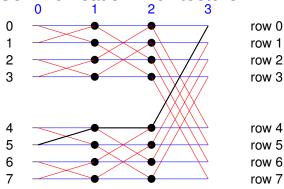


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Keep second bit. Blue (straight) edge.

Flip third bit. Red (cross edge).

Basis of communication switches!

Algorithms, circuits, and parallelism.

Add up n numbers; $a_0, \dots a_{n-1}$.

Algorithms, circuits, and parallelism.

Add up n numbers; $a_0, \dots a_{n-1}$.

O(n) time.

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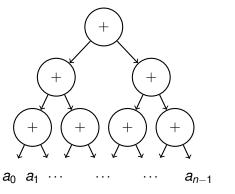
O(n) time.

Circuit:

Add up n numbers; $a_0, \dots a_{n-1}$.

O(n) time.

Circuit:

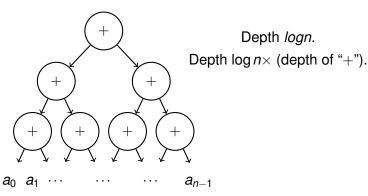


Depth logn.

Add up n numbers; $a_0, \dots a_{n-1}$.

O(n) time.

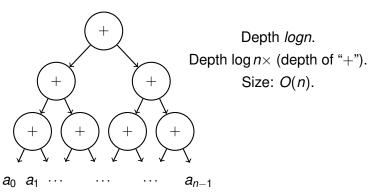
Circuit:



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O(n) time.

Circuit:



FFT Algorithm:

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Sequential: $O(n \log n)$

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How to build a circuit for Matrix Multiplication?

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Sequential: $O(n \log n)$

Depth: $O(\log n)$. Size: $O(n \log n)$.

How to build a circuit for Matrix Multiplication?

Who makes circuits anyway!!

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How to build a circuit for Matrix Multiplication?

Who makes circuits anyway!!

Not too many people.

Parallel?

Everywhere!

Shared memory model (SMM): Processors.

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"Ignores communication cost"

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Work: How much total work over processors.

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Circuits \Longrightarrow SMM

Time on SMM = Depth of circuit.

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Models parallelism.

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Algorithm:

Time: How many steps?

Work: How much total work over processors.

Circuits \Longrightarrow SMM

Time on SMM = Depth of circuit.

Work on SMM = Size of circuit

$$[a_0, a_1, a_2, \dots, a_{n-1}]$$

$$[a_0, a_1, a_2, \ldots, a_{n-1}]$$

Pair up and add.

$$[a_0, a_1, a_2, \dots, a_{n-1}]$$

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$$[(a_0+a_1),(a_2+a_3),\ldots,(a_{n-2}+a_{n-1})].$$

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Time: log n

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Time: log n

Work: $\frac{n}{2} + \frac{n}{4} + \cdots 1$

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And so on...

Time: log n

Work:
$$\frac{n}{2} + \frac{n}{4} + \cdots = O(n)$$
.

Algorithm: Split into two lists.

Algorithm: Split into two lists. Sort sublists

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Recursion depth: $O(\log n)$.

Algorithm:

Split into two lists.

Sort sublists

Merge sublists.

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Time: $O(\log n)$

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Merge was sequential!

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Make Merge faster?

Algorithm:

Split into two lists.

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Make Merge faster?

In $O(\log n)$ time using Mirrored FFT of comparators!

Bitonic Sort.

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Bitonic Sort.

Time: $O(\log^2 n)$.

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Algorithm:
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Time: $O(\log n)$??

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In $O(\log n)$ time using Mirrored FFT of comparators!

Bitonic Sort.

Time: $O(\log^2 n)$. Work: $O(n\log^2 n)$

Algorithm:

Split into two lists.

Sort sublists

Merge sublists.

Recursion depth: $O(\log n)$.

Time: $O(\log n)$??

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Time: $O(\log^2 n)$.

Work: $O(n\log^2 n)$ Worse than sequential!

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Cole's Mergesort:

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Bitonic Sort.

Time: $O(\log^2 n)$.

Work: $O(n\log^2 n)$ Worse than sequential!

Cole's Mergesort: $O(\log n)$ time;

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Merge sublists.

Recursion depth: $O(\log n)$.

Time: $O(\log n)$??

Merge was sequential!

Time: $O(n) + O(n/2) + O(n/4) \cdots = O(n)$.

Make Merge faster?

In $O(\log n)$ time using Mirrored FFT of comparators!

Bitonic Sort.

Time: $O(\log^2 n)$.

Work: $O(n\log^2 n)$ Worse than sequential!

Cole's Mergesort: $O(\log n)$ time; $O(n \log n)$ work.

```
Algorithm:
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Time: $O(\log n)$??

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In $O(\log n)$ time using Mirrored FFT of comparators!

Bitonic Sort.

Time: $O(\log^2 n)$.

Work: $O(n\log^2 n)$ Worse than sequential!

Cole's Mergesort: $O(\log n)$ time; $O(n \log n)$ work. Optimal!

Algorithm:

Split n bit numbers into n/2 bit numbers.

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Do three (or four) multiplications of those n/2 bit numbers.

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Recursion depth: log n.

Addition of *n*-bit numbers: O(n) time.

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Recursion depth: log *n*.

Addition of *n*-bit numbers: O(n) time.

Work mostly at the leaves:

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Shift and Add the numbers.

Recursion depth: log n.

Addition of *n*-bit numbers: O(n) time.

Work mostly at the leaves:

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Using the FFT!

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Show Some Code.

Multiplication on my

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