

Advanced Systems Lab

Spring 2020

Lecture: Optimizing FFT, FFTW

Instructor: Markus Püschel, Ce Zhang

TA: Joao Rivera, Bojan Karlas, several more



Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich

Fast FFT: Example FFTW Library

- www.fftw.org
- Frigo and Johnson, *FFTW: An Adaptive Software Architecture for the FFT*, ICASSP 1998
- Frigo, *A Fast Fourier Transform Compiler*, PLDI 1999
- Frigo and Johnson, *The Design and Implementation of FFTW3*, Proc. IEEE 93(2) 2005

Recursive Cooley-Tukey FFT

$$\text{DFT}_{km} = (\text{DFT}_k \otimes \text{I}_m) T_m^{km} (\text{I}_k \otimes \text{DFT}_m) L_k^{km}$$

radix

decimation-in-time

$$\text{DFT}_{km} = L_m^{km} (\text{I}_k \otimes \text{DFT}_m) T_m^{km} (\text{DFT}_k \otimes \text{I}_m)$$

decimation-in-frequency

- For powers of two $n = 2^t$ sufficient together with base case

$$\text{DFT}_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

3

Cooley-Tukey FFT, $n = 4$

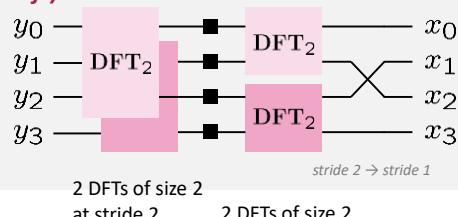
Fast Fourier transform (FFT)

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & i & -1 & -i \\ 1 & -1 & 1 & -1 \\ 1 & -i & -1 & i \end{bmatrix} = \begin{bmatrix} 1 & \cdot & 1 & \cdot \\ \cdot & 1 & \cdot & 1 \\ 1 & \cdot & -1 & \cdot \\ \cdot & 1 & \cdot & -1 \end{bmatrix} \begin{bmatrix} 1 & \cdot & \cdot & \cdot \\ \cdot & 1 & \cdot & \cdot \\ \cdot & \cdot & 1 & \cdot \\ \cdot & \cdot & \cdot & i \end{bmatrix} \begin{bmatrix} 1 & 1 & \cdot & \cdot \\ 1 & -1 & \cdot & \cdot \\ \cdot & \cdot & 1 & 1 \\ \cdot & \cdot & 1 & -1 \end{bmatrix} \begin{bmatrix} 1 & \cdot & \cdot & \cdot \\ \cdot & \cdot & 1 & \cdot \\ \cdot & 1 & \cdot & \cdot \\ \cdot & \cdot & \cdot & 1 \end{bmatrix}$$

Representation using matrix algebra

$$\text{DFT}_4 = (\text{DFT}_2 \otimes \text{I}_2) \text{ diag}(1, 1, 1, i) (\text{I}_2 \otimes \text{DFT}_2) L_2^4$$

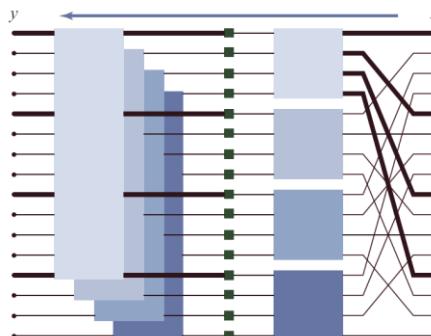
Data flow graph (right to left)



4

FFT, $n = 16$ (Recursive, Radix 4)

$$\text{DFT}_{16} = \begin{matrix} & \text{DFT}_4 \otimes I_4 & T_4^{16} & I_4 \otimes \text{DFT}_4 & L_4^{16} \\ \text{DFT}_{16} & = & \begin{array}{c} \text{Matrix} \\ \text{DFT}_4 \otimes I_4 \end{array} & \begin{array}{c} \text{Matrix} \\ T_4^{16} \end{array} & \begin{array}{c} \text{Matrix} \\ I_4 \otimes \text{DFT}_4 \end{array} & \begin{array}{c} \text{Matrix} \\ L_4^{16} \end{array} \end{matrix}$$



5

Fast Implementation (\approx FFTW 2.x)

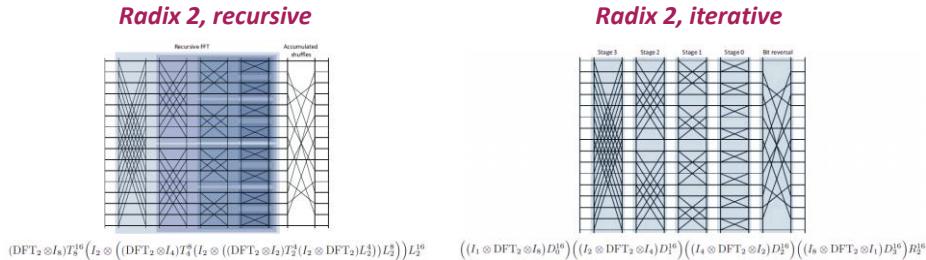
- Choice of algorithm
- Locality optimization
- Constants
- Fast basic blocks
- Adaptivity

6

1: Choice of Algorithm

- Choose recursive, not iterative

$$\text{DFT}_{km} = (\text{DFT}_k \otimes \text{I}_m) T_m^{km} (\text{I}_k \otimes \text{DFT}_m) L_k^{km}$$



First recursive implementation we consider in this course

7

2: Locality Improvement

$$\text{DFT}_{16} = \begin{matrix} \text{DFT}_4 \otimes I_4 & T_4^{16} & I_4 \otimes \text{DFT}_4 & L_4^{16} \end{matrix}$$

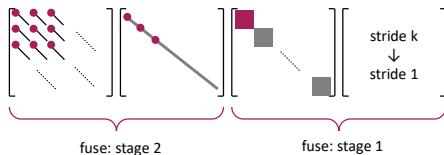
- Straightforward implementation: 4 steps
 - Permute
 - Loop recursively calling smaller DFTs (here: 4 of size 4)
 - Loop that scales by twiddle factors (diagonal elements of T)
 - Loop recursively calling smaller DFTs (here: 4 of size 4)
- 4 passes through data: bad locality
- Better: fuse some steps

8

2: Locality Improvement

$$DFT_n = (DFT_k \otimes I_m) T_m^n (I_k \otimes DFT_m) L_k^n$$

schematic:



- compute m many $DFT_k * D$ with input stride m and output stride m
- D is part of the diagonal T
- writes to the same location then it reads from \rightarrow inplace

- compute k many DFT_m with input stride k and output stride 1
- writes to different location then it reads from \rightarrow out-of-place

Interface needed for recursive call:

$DFTrec(m, x, y, k, 1);$

↑ ↑ ↑ ↑ ↑
 DFT size input/ output output input
 ↑ ↑ ↑ ↑ ↑
 input output vector stride stride

Can handle further recursion
(just strides change)

Interface needed for recursive call:

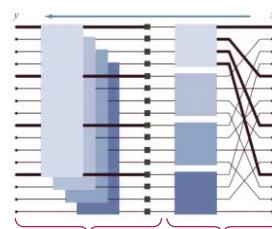
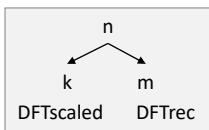
$DFTscaled(k, x, d, m);$

↑ ↑ ↑ ↑
 DFT size input = output diagonal elements
 ↑ ↑ ↑ ↑
 input output vector

Cannot handle further recursion so in FFTW it is a base case of the recursion

9

$$DFT_{km} = \underbrace{(DFT_k \otimes I_m)}_{\text{one loop}} T_m^{km} \underbrace{(I_k \otimes DFT_m)}_{\text{one loop}} L_k^{km}$$



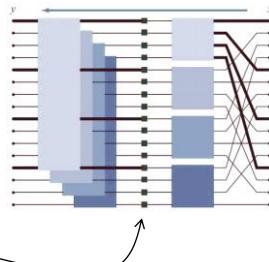
```
// code sketch
void DFT(int n, cpx *x, cpx *y) {
  ...
  int k = choose_dft_radix(n); // ensure k small enough
  int m = n/k;
  for (int i = 0; i < k; ++i)
    DFTrec(m, x + i, y + m*i, k, 1); // implemented as DFT(...) is
  for (int j = 0; j < m; ++j)
    DFTscaled(k, y + j, t[j], m); // always a base case
}
```

3: Constants

- FFT incurs multiplications by roots of unity
- In real arithmetic:
Multiplications by sines and cosines, e.g.,

```
y[i] = sin(i·pi/128)*x[i];
```
- Very expensive!
- **Observation:** Constants depend only on input size, not on input
- **Solution:** Precompute once and use many times

```
d = DFT_init(1024); // init function computes constant table  
d(x, y); // use many times
```



11

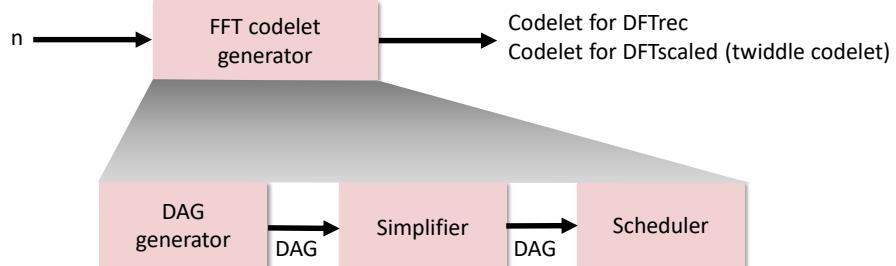
4: Optimized Basic Blocks

```
// code sketch  
void DFT(int n, cpx *x, cpx *y) {  
    if (use_base_case(n))  
        DFTbc(n, x, y); // use base case  
    else {  
        int k = choose_dft_radix(n); // ensure k <= 32  
        int m = n/k;  
        for (int i = 0; i < k; ++i)  
            DFTrec(m, x + i, y + m*i, k, 1); // implemented as DFT(...) is  
        for (int j = 0; j < m; ++j)  
            DFTscaled(k, y + j, t[j], m); // always a base case  
    }  
}
```

- Just like loops can be unrolled, recursions can also be unrolled
- Empirical study: Base cases for sizes $n \leq 32$ useful (scalar code)
- Needs 62 base cases or “codelets” (why?)
 - DFTrec, sizes 2–32
 - DFTscaled, sizes 2–32
- **Solution:** Codelet generator (codelet = optimized basic block)

12

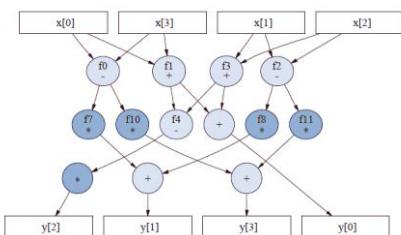
FFTW Codelet Generator



13

Small Example DAG

DAG:



One possible unparsing:

```

f0 = x[0] - x[3];
f1 = x[0] + x[3];
f2 = x[1] - x[2];
f3 = x[1] + x[2];
f4 = f1 - f3;
y[0] = f1 + f3;
y[2] = 0.7071067811865476 * f4;
f7 = 0.9238795325112867 * f0;
f8 = 0.3826834323650898 * f2;
y[1] = f7 + f8;
f10 = 0.3826834323650898 * f0;
f11 = (-0.9238795325112867) * f2;
y[3] = f10 + f11;

```

14

DAG Generator



- Knows FFTs: Cooley-Tukey, split-radix, Good-Thomas, Rader, represented in sum notation

$$y_{n_2j_1+j_2} = \sum_{k_1=0}^{n_1-1} (\omega_n^{j_2 k_1}) \left(\sum_{k_2=0}^{n_2-1} x_{n_1 k_2 + k_1} \omega_{n_2}^{j_2 k_2} \right) \omega_{n_1}^{j_1 k_1}$$

- For given n , suitable FFTs are recursively applied to yield n (real) expression trees for outputs y_0, \dots, y_{n-1}
- Trees are fused to an (unoptimized) DAG

15

Simplifier



- Applies:
 - Algebraic transformations
 - Common subexpression elimination (CSE)
 - DFT-specific optimizations
- Algebraic transformations
 - Simplify mults by 0, 1, -1
 - Distributivity law: $kx + ky = k(x + y)$, $kx + lx = (k + l)x$
Canonicalization: $(x-y), (y-x)$ to $(x-y), -(x-y)$
- CSE: standard
 - E.g., two occurrences of $2x+y$: assign new temporary variable
- DFT specific optimizations
 - All numeric constants are made positive (reduces register pressure)
 - CSE also on transposed DAG

16

Scheduler

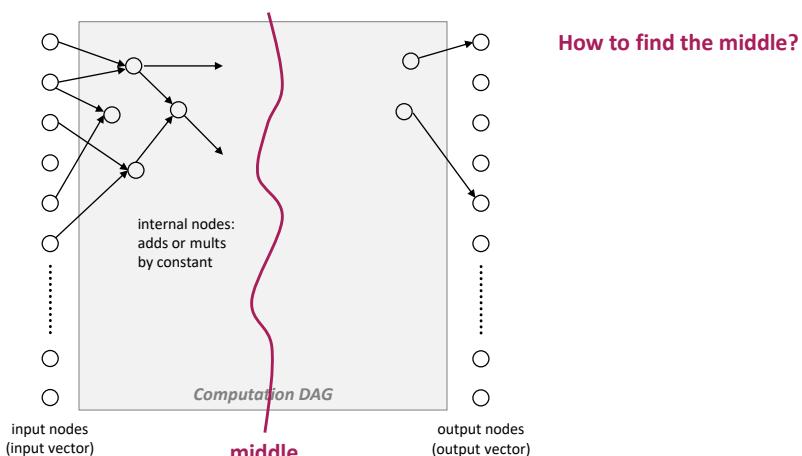


- Determines in which sequence the DAG is unparsed to C
(topological sort of the DAG)
Goal: minimizer register spills
- A 2-power FFT has an operational intensity of $I(n) = O(\log(C))$, where C is the cache size [1]
- Implies: For R registers $\Omega(n \log(n)/\log(R))$ register spills
- FFTW's scheduler achieves this (asymptotic) bound *independent* of R

[1] Hong and Kung: "[I/O Complexity: The red-blue pebbling game](#)"¹⁷

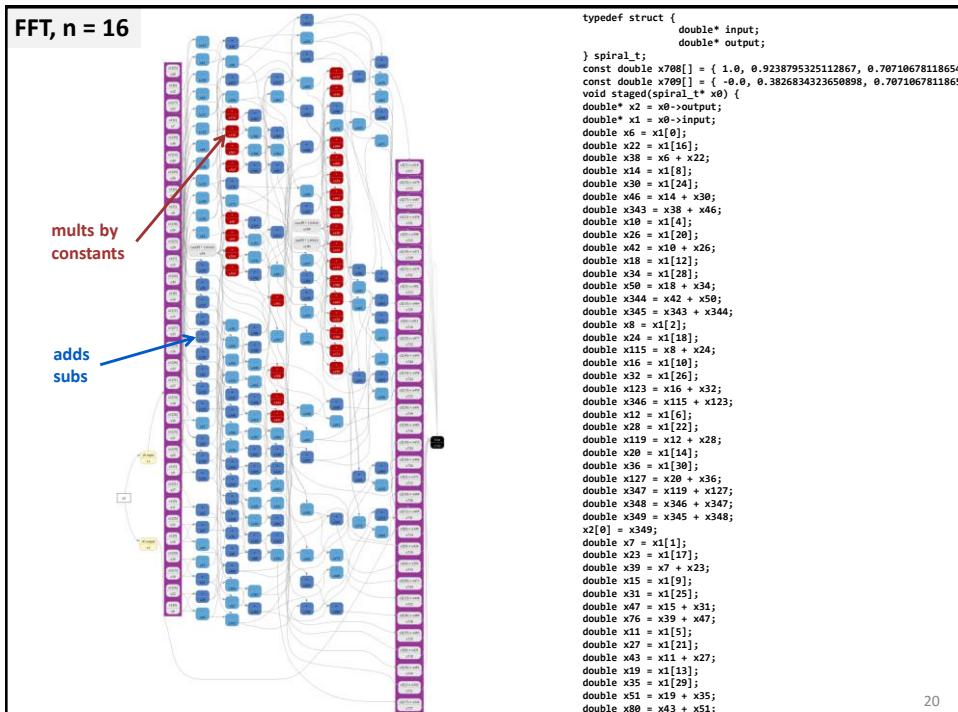
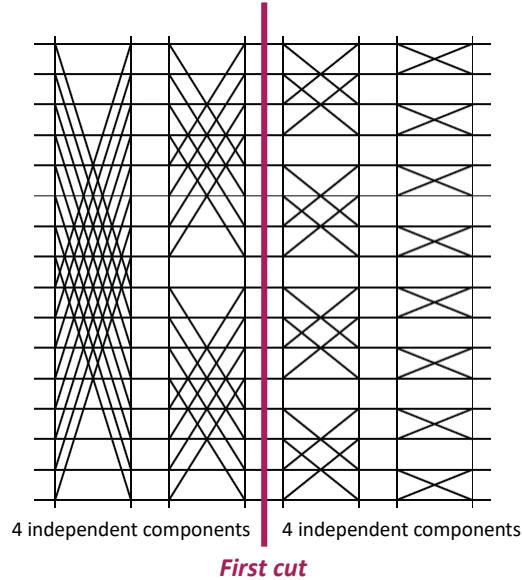
FFT-Specific Scheduler: Basic Idea

- Cut DAG in the middle
- Recurse on the connected components



18

This is a Sketched/Abstracted DAG



Codelet Examples

- [Notwiddle 2 \(DFTrec\)](#)
- [Notwiddle 3 \(DFTrec\)](#)
- [Twiddle 3 \(DFTscaled\)](#)
- [Notwiddle 32 \(DFTrec\)](#)

- **Code style:**
 - Single static assignment (SSA)
 - Scoping (limited scope where variables are defined)

21

5: Adaptivity

Choices used for platform adaptation

```
// code sketch
void DFT(int n, cpx *x, cpx *y) {
    if (use_base_case(n)) {
        DFTbc(n, x, y); // use base case
    } else {
        int k = choose_dft_radix(n); // ensure k <= 32
        int m = n/k;
        for (int i = 0; i < k; ++i)
            DFTrec(m, x + i, y + m*i, k, 1); // implemented as DFT(...) is
        for (int j = 0; j < m; ++j)
            DFTscaled(k, y + j, t[j], m); // always a base case
    }
}
```

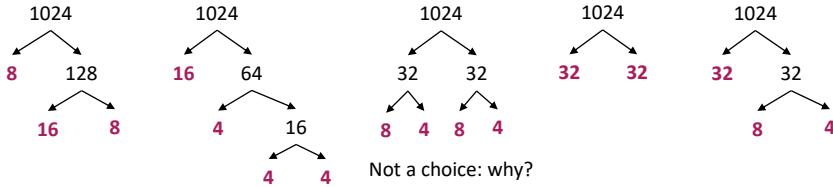
```
d = DFT_init(1024); // compute constant table; search for best recursion
d(x, y);           // use many times
```

22

5: Adaptivity

```
d = DFT_init(1024); // compute constant table; search for best recursion  
d(x, y); // use many times
```

Choices: $DFT_{km} = (DFT_k \otimes I_m)T_m^{km}(I_k \otimes DFT_m)L_k^{km}$



Base case = generated codelet is called

- Exhaustive search is expensive
- Solution: Dynamic programming

23

FFTW: Further Information

- Previous Explanation: FFTW 2.x
- FFTW 3.x:
 - Support for SIMD/threading
 - Flexible interface to handle FFT variants (real/complex, strided access, sine/cosine transforms)
 - Complicates significantly the interfaces actually used and increases the size of the search space

24

	MMM <i>Atlas</i>	Sparse MVM <i>Sparsity/Bebop</i>	DFT <i>FFTW</i>
Cache optimization			
Register optimization			
Optimized basic blocks			
Other optimizations			
Adaptivity			

	MMM <i>Atlas</i>	Sparse MVM <i>Sparsity/Bebop</i>	DFT <i>FFTW</i>
Cache optimization	Blocking	Blocking (rarely useful)	Recursive FFT, fusion of steps
Register optimization	Blocking	Blocking (changes sparse format)	Scheduling of small FFTs
Optimized basic blocks	Unrolling, scalar replacement and SSA, scheduling, simplifications (for FFT)		
Other optimizations	—	—	Precomputation of constants
Adaptivity	Search: blocking parameters	Search: register blocking size	Search: recursion strategy