Algorithms and Computation in Signal Processing

special topic course 18-799B spring 2005 24th and 25th Lecture Apr. 07 and 12, 2005

Instructor: Markus Pueschel TA: Srinivas Chellappa

Research Projects

Presentations last week of April (26th and 28th)

- We distribute the dates in the lecture on the 12th
- Presentations 20 minutes + 5 minutes questions (~17-20 slides)

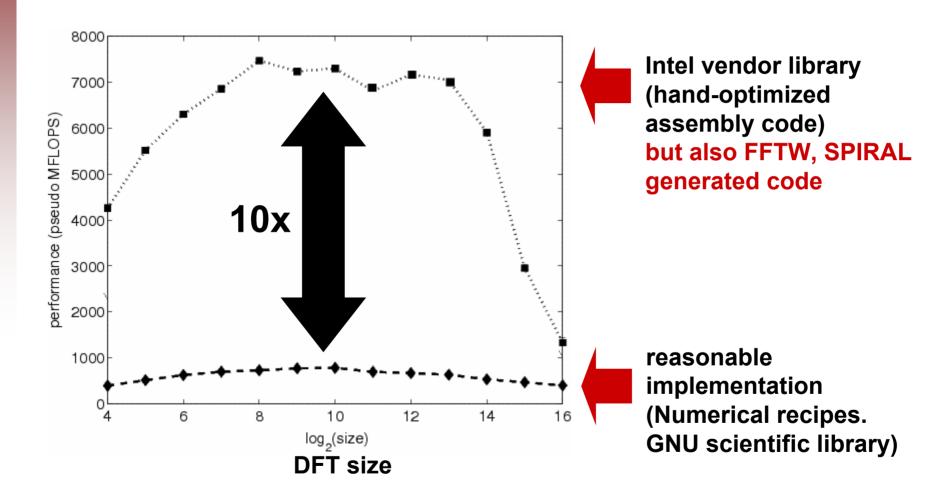
Research paper

- Due April 20th, the only thing that may be missing are some (but not all) experimental results
- You'll get feedback from me
- Final version with feedback incorporated due one week after your presentation

Remarks

- Follow guide to benchmarking!
- Try different sets of compiler flags to be sure
- Do a cost analysis

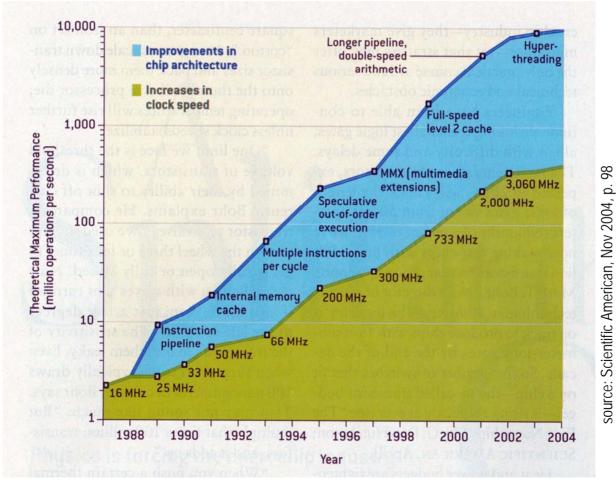
The Problem Again



Writing fast numerical code is a tough problem

Moore's Law

Moore's Law: exponential (x2 in ~18 months) increase number of transistors/chip



But everything has its price ...

Moore's Law: Consequences

Computers are very complex

- multilevel memory hierarchy
- special instruction sets beyond standard C programming model
- undocumented hardware optimizations

Consequences:

- Runtime depends only roughly on the operations
- Runtime behavior is hard to understand
- Compiler development can hardly keep track
- The best code (and algorithm) is platform-dependent
- It is very difficult to write really fast code

Computers evolve fast

- Highly tuned code becomes obsolete almost as fast as it as written
- It'll get rather worse: Multicoresystems

Solution #1: Brute Force

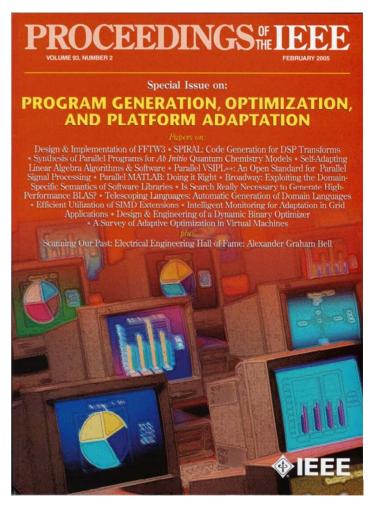
Thousands of programmers hand-write and hand-tune (assembly) code for the same numerical problems and for every platform and whenever a new platform comes out?

Hmm.....

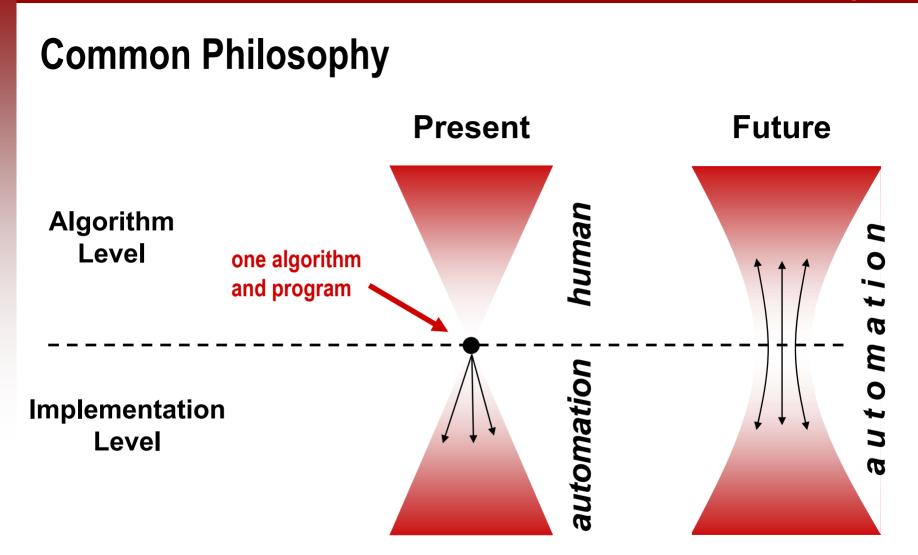
(but it's current practice)

Solution #2: New Approaches to Code Optimization and Code Creation

- ATLAS: Code generation/optimization for BLAS
- SPARSITY/BeBop: Code generation/optimization for sparse linear algebra routines
- FFTW: Self-adaptive DFT library + DFT kernel generator
 - **SPIRAL:** Code generation/optimization for linear signal transforms



Proceedings of the IEEE special issue, Feb. 2005



a new breed of domain-aware approaches/tools push automation beyond what is currently possible applies for software and hardware design alike

SPIRAL www.spiral.net

Sponsors:

DARPA NSF ACR-0234293 NSF ITR/NGS-0325687

and:

Cylab, CMU Austrian Science Fund Intel ITRI, Taiwan ENSCO, Inc.

~40 Publications

Overview paper:

Markus Püschel, José M. F. Moura, Jeremy Johnson, David Padua, Manuela Veloso, Bryan Singer, Jianxin Xiong, Franz Franchetti, Aca Gacic, Yevgen Voronenko, Kang Chen, Robert W. Johnson, and Nick Rizzolo, **SPIRAL: Code Generation for DSP Transforms,** Proceedings of the IEEE

Team:

James C. Hoe (ECE, CMU) Jeremy Johnson (CS, Drexel) José M. F. Moura (ECE, CMU) David Padua (CS, UIUC) Markus Püschel (ECE, CMU) Manuela Veloso (CS, CMU) Robert W. Johnson (Quarry Comp. Inc.) Christoph Überhuber (Math, TU Wien

Students/PostDocs:

Bryan W. Singer (CS, CMU) Jianxin Xiong (CS, UIUC) Srinivas Chellappa (ECE, CMU) Franz Franchetti (ECE, CMU, before TU Vienna) Aca Gacic (ECE, CMU) Yevgen Voronenko (ECE, CMU) Anthony Breitzman (CS, Drexel) Kang Chen (CS, Drexel) Pinit Kumhom (ECE, Drexel) Adam Zelinski (ECE, CMU) Peter Tummeltshammer (CS, TU Vienna)

Spiral

- Code generation from scratch for linear digital signal processing (DSP) transforms (DFT, DCT, DWT, filters,)
- Automatic optimization and platform-tuning at the algorithmic level and the code level
- Different code types supported (scalar, vector, FMA, fixedpoint, multiplierless, ...)
- Goal: A <u>flexible</u>, <u>extensible</u> code generation framework that can <u>survive time</u> (to whatever extent possible) for an entire domain of algorithms

Research question: To what extent is it possible to abolish handcoding and handoptimization?

Code Generation and Tuning as Optimization Problem

- T a DSP transform to be implemented
- P the target platform
- $\mathcal{I} = \mathcal{I}(\mathbf{T}, \mathbf{P})$ set of possible implementations of T on P
- $C=C(T,I,P)\;$ cost of implementation I of T on P

The implementation of T that is tuned to P is given by:

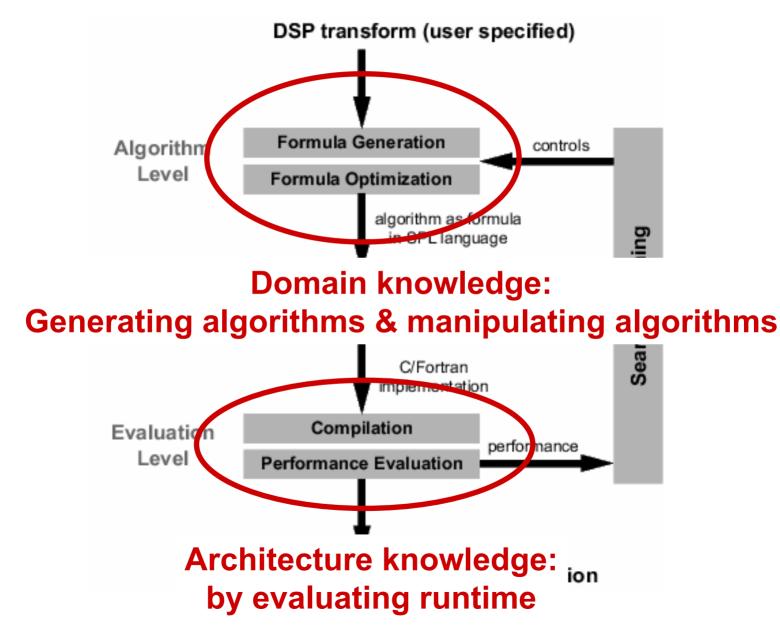
$$\widehat{\mathbf{I}} = \widehat{\mathbf{I}}(\mathbf{P}) = \operatorname{arg\,min}_{\mathbf{I} \in \mathcal{I}(\mathbf{P})} \mathbf{C}(\mathbf{T}, \mathbf{P}, \mathbf{I})$$

Problems:

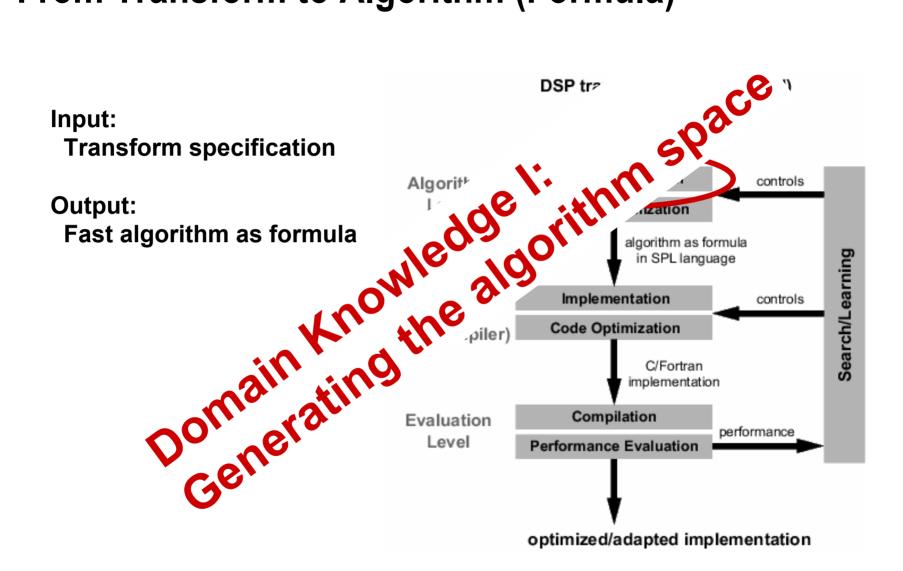
- How to characterize and generate the set of implementations?
- How to efficiently minimize C?

Spiral exploits the <u>domain-specific structure</u> to implement a solver for this optimization problem

Spiral's architecture



From Transform to Algorithm (Formula)



DSP Algorithms: Example 4-point DFT

Cooley/Tukey FFT (size 4):

 $\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & i & -1 & -i \\ 1 & -1 & 1 & -1 \\ 1 & -i & -1 & i \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & i \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & -1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$

Fourier transformDiagonal matrix (twiddles)II $DFT_4 = (DFT_2 \otimes I_2) \cdot T_2^4 \cdot (I_2 \otimes DFT_2) \cdot L_2^4$ IIIIKronecker productIdentityPermutation

- mathematical notation exhibits structure: SPL (signal processing language)
- Suitable for computer representation
- contains all information to generate code

SPL: Definition (BNF)

Description language for linear DSP algorithms Definition (BNF):

(product) (direct sum) (tensor product) (overlapped tensor product) (conversion to real)

Some Definitions:

$$A \oplus B = \begin{bmatrix} A \\ B \end{bmatrix}$$
$$A \otimes B = \begin{bmatrix} a_{k,\ell}B \end{bmatrix}, \text{ where } A = \begin{bmatrix} a_{k,\ell} \end{bmatrix} \qquad I_n \otimes A = \begin{bmatrix} A \\ A \\ \vdots \\ F_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$
$$\mathsf{F}_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \qquad \mathsf{R}_\alpha = \begin{bmatrix} \cos \alpha & \sin \alpha \\ -\sin \alpha & \cos \alpha \end{bmatrix}$$

DSP Algorithms: Spiral Terminology

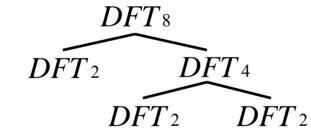
Transform *DFT_n* parameterized matrix

Rule $DFT_{nm} \rightarrow (R)$

$$DFT_{nm} \rightarrow (DFT_n \otimes I_m) \cdot D \cdot (I_n \otimes DFT_m) \cdot P$$

a breakdown strategy product of sparse matrices

Ruletree



- recursive application of rules
- uniquely defines an algorithm
- efficient representation
- easy manipulation

Formula

$$DFT_8 = (F_2 \otimes I_4) \cdot D \cdot (I_2 \otimes (I_2 \otimes F_2 \cdots)) \cdot P$$

- few constructs and primitives
- uniquely defines an algorithm
- can be translated into code

Some Transforms

$$\begin{aligned} \mathbf{D}\mathbf{C}\mathbf{T}\mathbf{-2}_{n} &= \left[\cos(k(2\ell+1)\pi/2n)\right]_{0\leq k,\ell < n},\\ \mathbf{D}\mathbf{C}\mathbf{T}\mathbf{-3}_{n} &= \mathbf{D}\mathbf{C}\mathbf{T}\mathbf{-2}_{n}^{T} \quad (\text{transpose}),\\ \mathbf{D}\mathbf{C}\mathbf{T}\mathbf{-4}_{n} &= \left[\cos((2k+1)(2\ell+1)\pi/4n)\right]_{0\leq k,\ell < n},\\ \mathbf{I}\mathbf{M}\mathbf{D}\mathbf{C}\mathbf{T}_{n} &= \left[\cos((2k+1)(2\ell+1+n)\pi/4n)\right]_{0\leq k<2n,0\leq \ell < n},\\ \mathbf{R}\mathbf{D}\mathbf{F}\mathbf{T}_{n} &= \left[r_{k\ell}\right]_{0\leq k,\ell < n}, \quad r_{k\ell} = \begin{cases} \cos\frac{2\pi k\ell}{n}, \quad k\leq \lfloor\frac{n}{2}\rfloor\\ -\sin\frac{2\pi k\ell}{n}, \quad k> \lfloor\frac{n}{2}\rfloor,\\ -\sin\frac{2\pi k\ell}{n}, \quad k> \lfloor\frac{n}{2}\rfloor,\\ \end{bmatrix}\\ \mathbf{W}\mathbf{H}\mathbf{T}_{n} &= \begin{bmatrix} \mathbf{W}\mathbf{H}\mathbf{T}_{n/2} \quad \mathbf{W}\mathbf{H}\mathbf{T}_{n/2}\\ \mathbf{W}\mathbf{H}\mathbf{T}_{n/2} - \mathbf{W}\mathbf{H}\mathbf{T}_{n/2} \end{bmatrix}, \quad \mathbf{W}\mathbf{H}\mathbf{T}_{2} = \mathbf{D}\mathbf{F}\mathbf{T}_{2},\\ \mathbf{D}\mathbf{H}\mathbf{T} &= \left[\cos(2k\ell\pi/n) + \sin(2k\ell\pi/n)\right]_{0\leq k,\ell < n}.\end{aligned}$$

Spiral currently contains 36 transforms

Some Breakdown Rules

$$\begin{array}{rcl} \mathrm{DFT}_n & \rightarrow & (\mathrm{DFT}_k \otimes \mathrm{I}_m) \ \mathsf{T}_m^n(\mathrm{I}_k \otimes \mathrm{DFT}_m) \ \mathsf{L}_k^n, & n = km \\ \mathrm{DFT}_n & \rightarrow & P_n(\mathrm{DFT}_k \otimes \mathrm{DFT}_m) Q_n, & n = km, \ \gcd(k,m) = 1 \\ \mathrm{DFT}_p & \rightarrow & R_p^T(\mathrm{I}_1 \oplus \mathrm{DFT}_{p-1}) D_p(\mathrm{I}_1 \oplus \mathrm{DFT}_{p-1}) R_p, & p \ \text{prime} \\ \mathrm{DCT-3}_n & \rightarrow & (\mathrm{I}_m \oplus \mathrm{J}_m) \ \mathsf{L}_m^n(\mathrm{DCT-3}_m(1/4) \oplus \mathrm{DCT-3}_m(3/4)) \\ & & \cdot(\mathsf{F}_2 \otimes \mathrm{I}_m) \begin{bmatrix} \mathrm{I}_m & 0 \oplus - \mathrm{J}_{m-1} \\ \frac{1}{\sqrt{2}}(\mathrm{I}_1 \oplus 2\mathrm{I}_m) \end{bmatrix}, & n = 2m \\ \mathrm{DCT-4}_n & \rightarrow & S_n\mathrm{DCT-2}_n \ \mathrm{diag}_{0 \leq k < n}(1/(2\cos((2k+1)\pi/4n)))) \\ \mathrm{IMDCT}_{2m} & \rightarrow & (\mathrm{J}_m \oplus \mathrm{I}_m \oplus \mathrm{J}_m) \left(\left(\begin{bmatrix} 1 \\ -1 \end{bmatrix} \otimes \mathrm{I}_m \right) \oplus \left(\begin{bmatrix} -1 \\ -1 \end{bmatrix} \otimes \mathrm{I}_m \right) \right) \ \mathrm{J}_{2m} \ \mathrm{DCT-4}_{2m} \\ \mathrm{WHT}_{2k} & \rightarrow & \prod_{i=1}^t (\mathrm{I}_{2k_1 + \dots + k_{i-1}} \otimes \mathrm{WHT}_{2^{k_i}} \otimes \mathrm{I}_{2^{k_{i+1} + \dots + k_t}), & k = k_1 + \dots + k_t \\ \mathrm{DFT}_2 & \rightarrow & \mathsf{F}_2 \\ \mathrm{DCT-2}_2 & \rightarrow & \mathrm{diag}(1, 1/\sqrt{2}) \ \mathsf{F}_2 \\ \mathrm{DCT-4}_2 & \rightarrow & \mathrm{J}_2 \ \mathsf{R}_{13\pi/8} \end{array}$$

Spiral contains 100+ rules

Some Breakdown Rules for Filters

$$\begin{aligned} \operatorname{Filt}_{n}(h(z)) \to \operatorname{I}_{\lfloor \frac{n}{b} \rfloor} \otimes_{l+r} \left(\begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix} \right) & \oplus^{k} \operatorname{T}_{k} \left(h(z) z^{l-\lceil \frac{l+r}{b} \rceil b-k} \right) \right) \\ \operatorname{Filt}_{n}(h(z)) \to \operatorname{L}_{\frac{n}{2}}^{n} \operatorname{Filt}_{\frac{n}{2}} \left(\begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix} \right) & \cdot \\ \operatorname{Filt}_{\frac{n}{2}} \left(\begin{bmatrix} h_{0}(z) & & \\ & h_{1}(z) & \\ & & h_{0}(z) + h_{1}(z) \end{bmatrix} \right) & \operatorname{Filt}_{\frac{n}{2} + \frac{r+l-1}{2}} \left(\begin{bmatrix} 1 & -1 \\ z & -1 \\ 0 & 1 \end{bmatrix} \right) & \cdot \operatorname{L}_{2}^{n+r+l} \end{aligned}$$

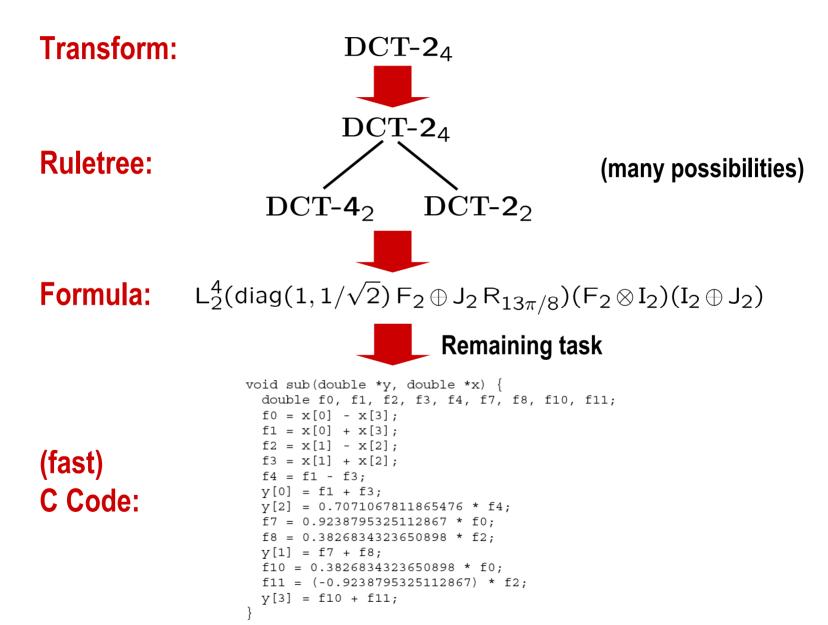
 $\operatorname{Filt}_n(h(z)) \to \operatorname{R}_{n,l,r}^{\operatorname{zero}} \cdot \operatorname{C}_{n+l+r}(h(z))$

$$\mathbf{C}_n(h(z)) \to \mathbf{RDFT}_n^{-1} \cdot X(\hat{\mathbf{h}}) \cdot \mathbf{RDFT}_n,$$

 $\hat{\mathbf{h}} = \mathbf{RDFT}_n \cdot \mathbf{h}$

Aca Gacic, Automatic Implementation and Platform Adaptation of Discrete Filtering and Wavelet Algorithms, Ph.D. thesis, Electrical and Computer Engineering, Carnegie Mellon University, 2004

Formula- (Algorithm) generation



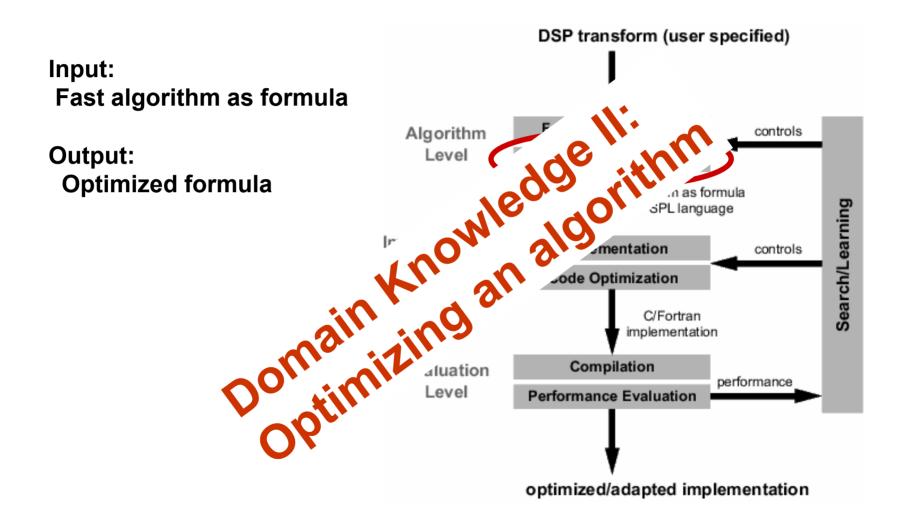
Set of Algorithms

Given a transform:

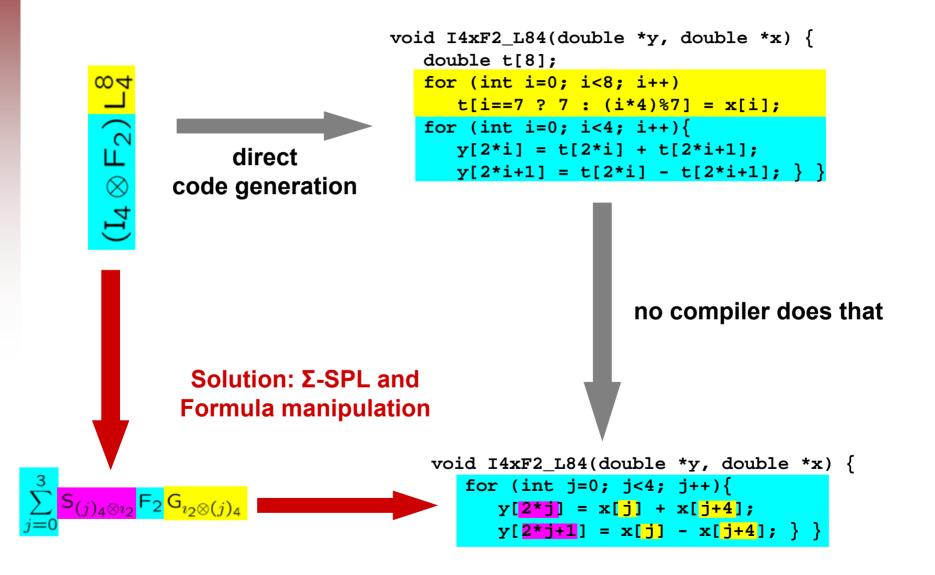
- Apply breakdown rules recursively until all occurring transforms are expanded
- Choice of rules at each step yields (usually) exponentially large algorithms space:
 - about equal in operations count
 - differ in data flow

| k | # DFTs, size 2 ^k | # DCT IV, size 2 ^k |
|---|-----------------------------|-------------------------------|
| 1 | 1 | 1 |
| 2 | 6 | 10 |
| 3 | 40 | 126 |
| 4 | 296 | 31242 |
| 5 | 27744 | 1924443362 |
| 6 | 162570361280 | 7343815121631354242 |
| 7 | ~1.01 • 10^27 | ~1.07 • 10^38 |
| 8 | ~2.31 • 10^61 | ~2.30 • 10^76 |
| 9 | ~2.86 • 10^133 | ~1.06 • 10^153 |

From Algorithm (Formula) to Optimized Algorithm



Motivation: Loop Fusion

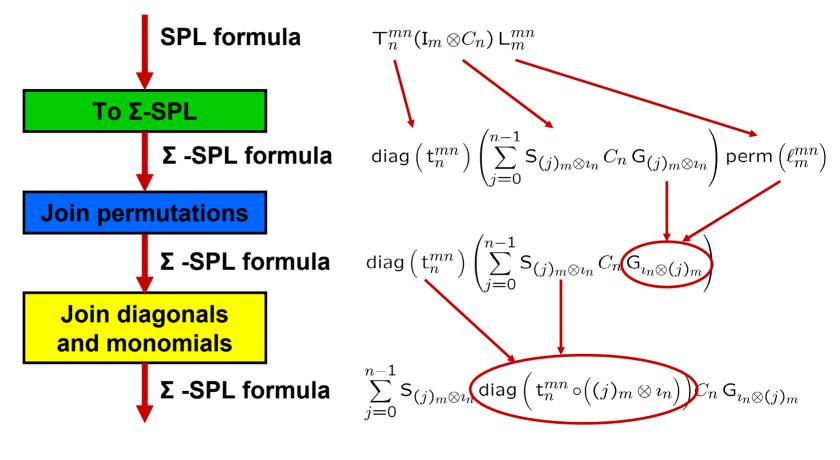


Formula Level Optimization

Main goals:

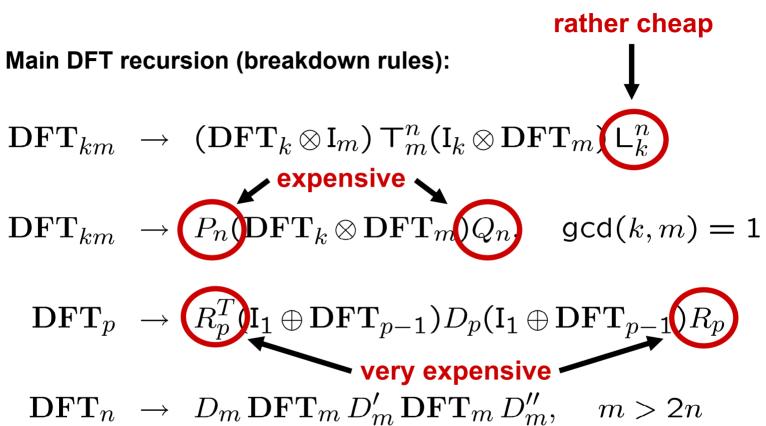
- Fusing iterative steps (fusing loops), e.g., permutations with loops
- Improving structure (data flow) for SIMD instructions
- Overcomes compiler limitations
- Formula manipulation through mathematical rules
- Implemented using multiple levels of rewriting systems
- Puts math knowledge into the system

Structure of Loop Optimization



Rules: $G_r \operatorname{perm}(\pi) = G_{\pi \circ r}, \quad \ell_m^{mn} \circ ((j)_m \otimes \iota_n) = \iota_n \otimes (j)_m$ $\operatorname{diag}(f) S_w = S_w \operatorname{diag}(f \circ w)$

Loop Fusion Beyond Cooley-Tukey



How to fuse permutations from different combinations of rules?

Example

array

Consider the DFT formula fragment

$$(I_p \otimes (I_1 \oplus (I_r \otimes DFT_s)L_r^{rs}) W_p) V_{p,q}$$

$$Cooley-Tukey Rader Good-Thomas$$

$$In \Sigma-SPL: \sum_{j_1=0}^{p-1} \left(S_{((j_1)_p \otimes i_q) \circ (0)_+^{1 \to q} \circ i_1} G_{v^{p,q} \circ ((j_1)_p \otimes i_q) \circ \overline{w}_{1,g}^q \circ (0)_+^{1 \to q}} + \sum_{j_0=0}^{r-1} S_{((j_1)_p \otimes i_q) \circ (1)_+^{q-1 \to q} \circ ((j_0)_r \otimes i_s)} DFT_s$$

$$Complicated array access G_{v^{p,q} \circ ((j_1)_p \otimes i_q) \circ \overline{w}_{1,g}^q \circ (1)_+^{q-1 \to q} \circ \ell_r^{rs} \circ ((j_0)_r \otimes i_s)} \right).$$

Example (cont'd)

After index function simplification:

$$\sum_{\substack{j_1=0\\b_1=qj_1}}^{p-1} \left(S_{h_{0,q}^{p\to pq} \circ (j_1)_p} G_{\hbar_{0,q}^{p\to pq} \circ (j_1)_p} + \sum_{\substack{j_0=0\\\phi_1=g^{j_0}}}^{r-1} S_{h_{qj_1+sj_0+1,1}^{s\to pq}} \operatorname{DFT}_s G_{\hbar_{b_1,p}^{q\to pq} \circ w_{\phi_1,g^s}} \right)$$

Simplified

array access

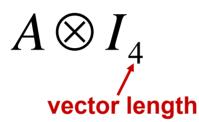
Example (cont'd)

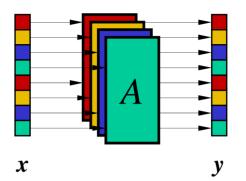
Generated Code

```
// Input: _Complex double x[28], output: y[28]
int p1, b1;
for(int j1 = 0; j1 <= 3; j1++) {
   y[7*j1] = x[(7*j1\%28)];
   p1 = 1; b1 = 7*j1;
   for(int j0 = 0; j0 <= 2; j0++) {
      y[b1 + 2*j0 + 1] =
         x[(b1 + 4*p1)%28] + x[(b1 + 24*p1)%28];
      y[b1 + 2*j0 + 2] =
         x[(b1 + 4*p1)%28] - x[(b1 + 24*p1)%28];
      p1 = (p1*3\%7);
   }
}
```

Vector code generation from SPL formulas

Naturally vectorizable construct





(Current) generic construct completely vectorizable:

$$\prod_{i=1}^{k} P_i D_i (A_i \otimes I_{\upsilon}) E_i Q_i \qquad \begin{array}{c} P_{\flat} Q_i & \text{permutations} \\ D_{\flat} E_i & \text{diagonals} \\ A_i & \text{arbitrary formulas} \\ \nu & \text{SIMD vector length} \end{array}$$

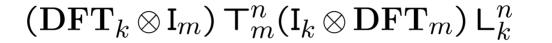
Vectorization in two steps:

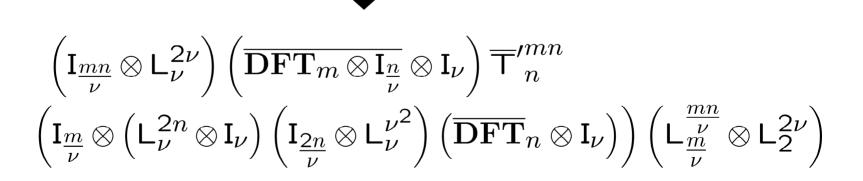
- 1. Formula manipulation using manipulation rules
- 2. Code generation (vector code + C code)

Formula manipulation overcomes compiler limitations



Standard FFT





Formula manipulation

Vector FFT for v-way vector instructions

Implementation of Formula Generation and Manipulation

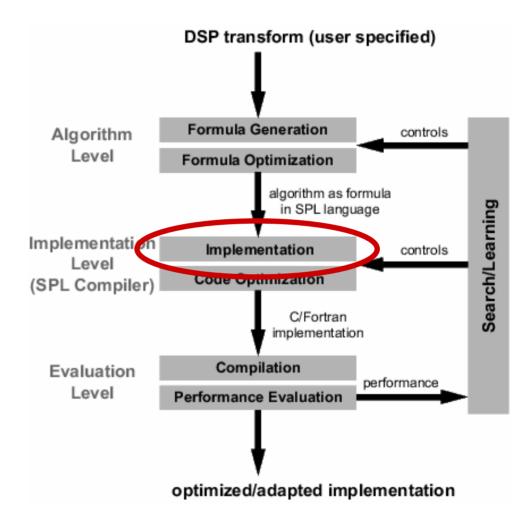
- Implementation using a computer algebra system (GAP)
- SPL/Σ-SPL implemented as recursive data types
- Exact representation of sin(), cos(), etc.
- Symbolic computation enables exact verification of rules

From Optimized Algorithm (Formula) to Code

Input: Optimized formula

Output: Intermediate Code

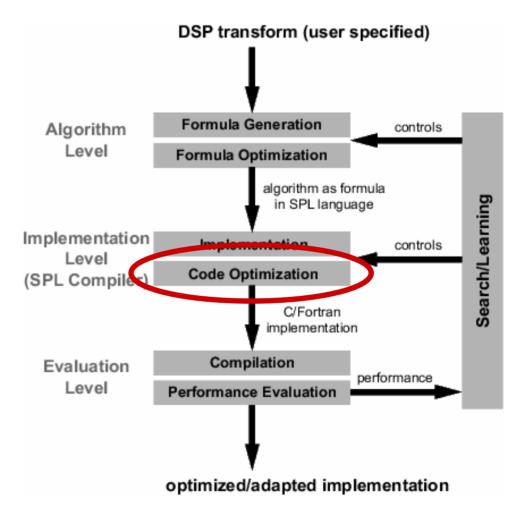
Straightforward



From Code to Optimized Code

Input: Intermediate Code

Output: Optimized C code



Code Level Optimizations

- Precomputation of constants
- Loop unrolling (controlled by search module)
- Constant inlining
- SSA code, scalar replacement, algebraic simplifications, CSE
- Code reordering for locality (optional)
- Conversion to FMA code (optional)
- Conversion to fixed point code (optional)
- Conversion to multiplierless code (optional)
- Finally: Unparsing to C (or Fortran)

Conversion to FMA code

- FMA (fused multiply-add) or MAC (multiply accumulate) instructions: y = ±ax ± b
- Extension of the instruction set + specialized execution units
- As fast as a single add or multiply
- Conversion of linear algorithms to FMA code: blackboard
- Paper: Yevgen Voronenko and Markus Püschel Automatic Generation of Implementations for DSP Transforms on Fused Multiply-Add Architectures Proc. (ICASSP) 2004

Evaluating Code

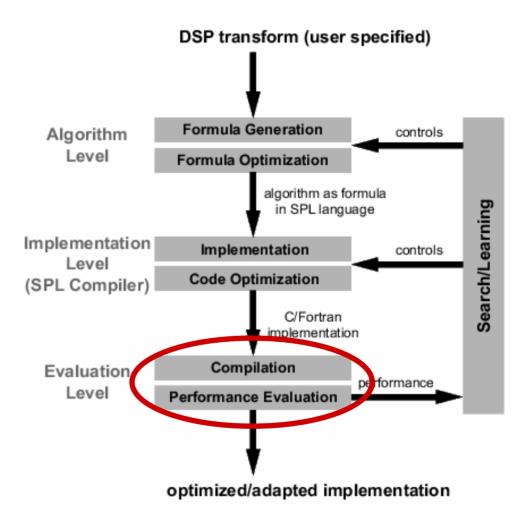
Input: Optimized C code

Output: Performance Number

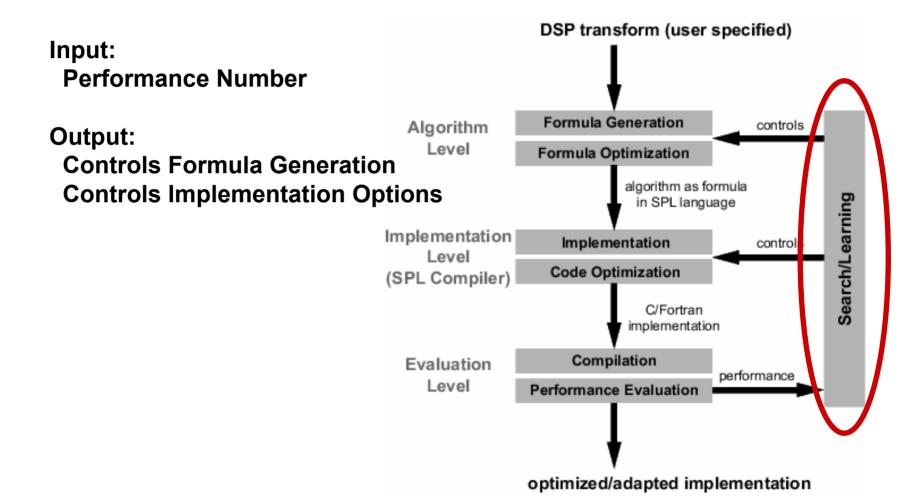
Straightforward

Examples:

- runtime
- accuracy
- operations count



Search (Learning) for the Best



Search Methods

Search over:

- Algorithmic degrees of freedom (choice of breakdown rules)
- Implementation degrees of freedom (degree of unrolling)

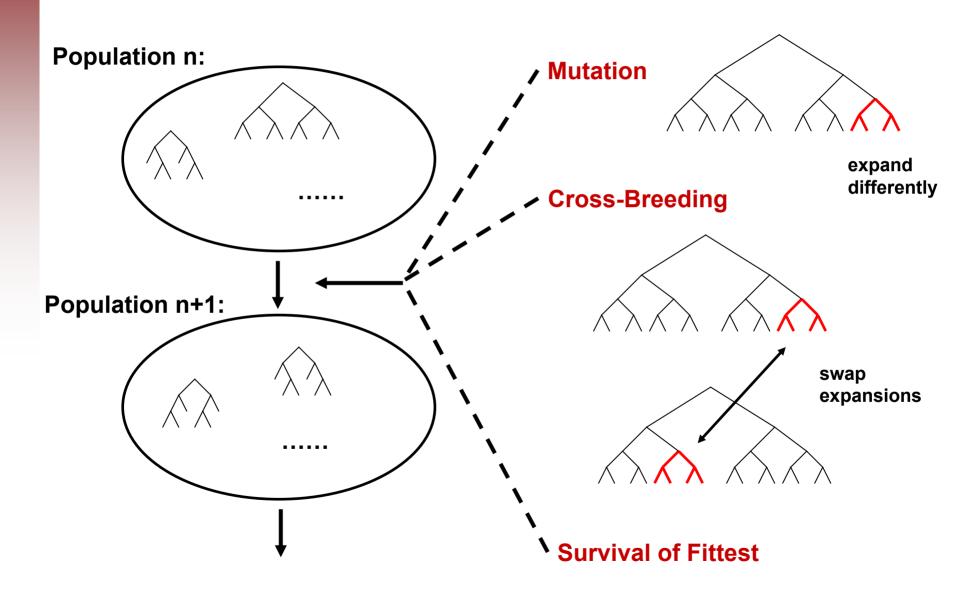
Operates with the ruletree representation of an algorithm

- transform independent
- efficient

Search Methods

- Exhaustive Search
- Dynamic Programming (DP)
- Random Search
- Hill Climbing
- STEER (an evolutionary algorithm)

STEER: Evolutionary Search



Learning

Procedure:

- Generate a set of (1000 say) algorithms and their runtimes (one transform, one size); represent algorithms by features
- From this data (pairs of features and runtimes), learn a set of algorithm design rules
- From this set, generate best algorithms (theory of Markov decision processes)

Evaluation:

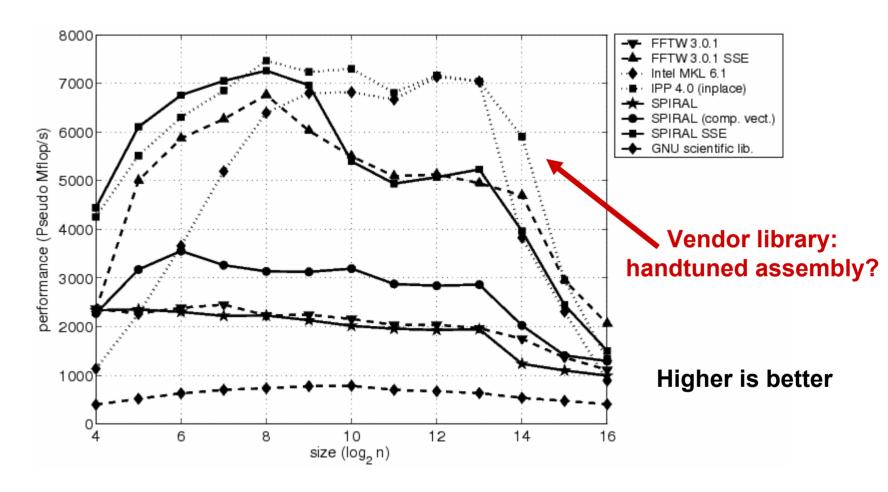
- Tested for WHT and DFT
- From data generated for one size (2¹⁵) could construct best algorithms across sizes (2¹²⁻²18)

Bryan Singer and Manuela Veloso Learning to Construct Fast Signal Processing Implementations Journal of Machine Learning Research, 2002, Vol. 3, pp. 887-919

Benchmarks

Benchmark: DFT, 2-powers

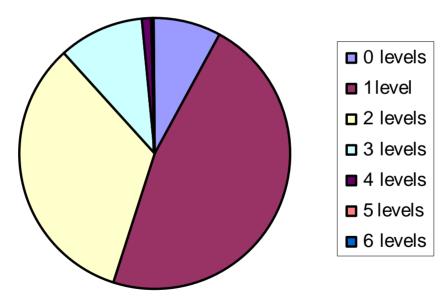
P4, 3.2 GHz, icc 8.0



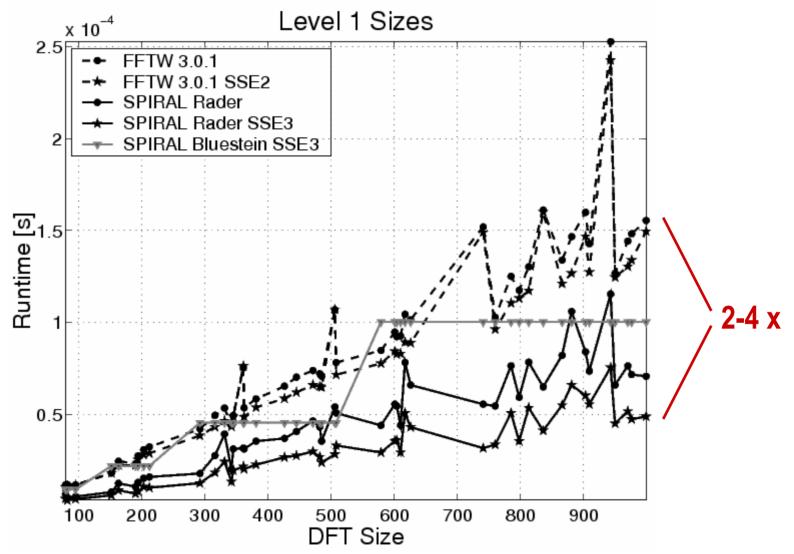
Single precision

Benchmark: DFT, Other Sizes

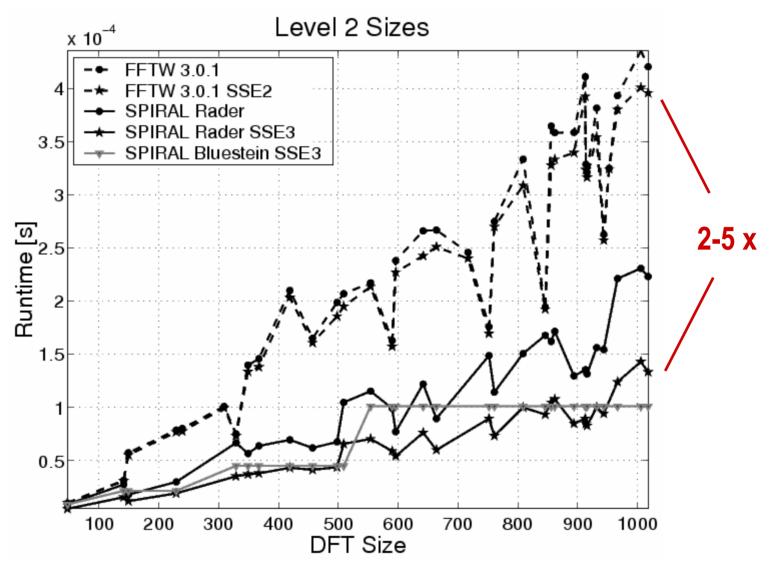
Divide sizes into levels by number of necessary Rader steps
 n < 8192



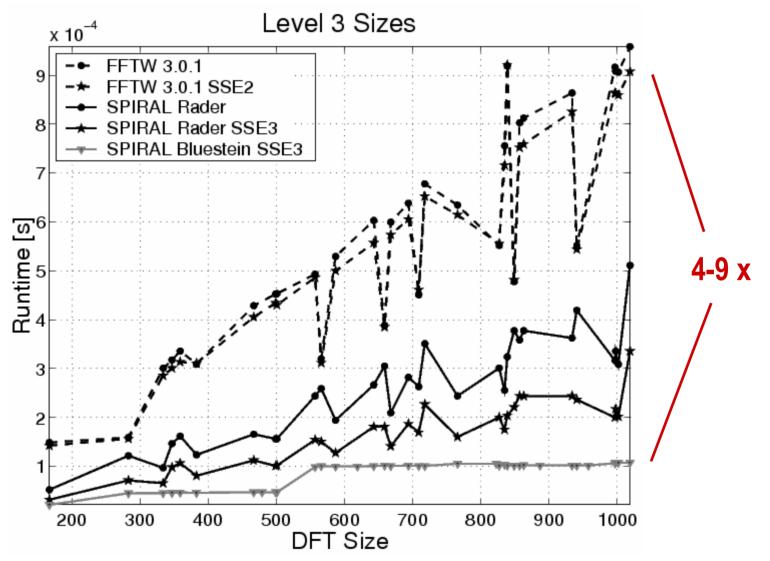
Benchmark: DFT, Level 1 Sizes



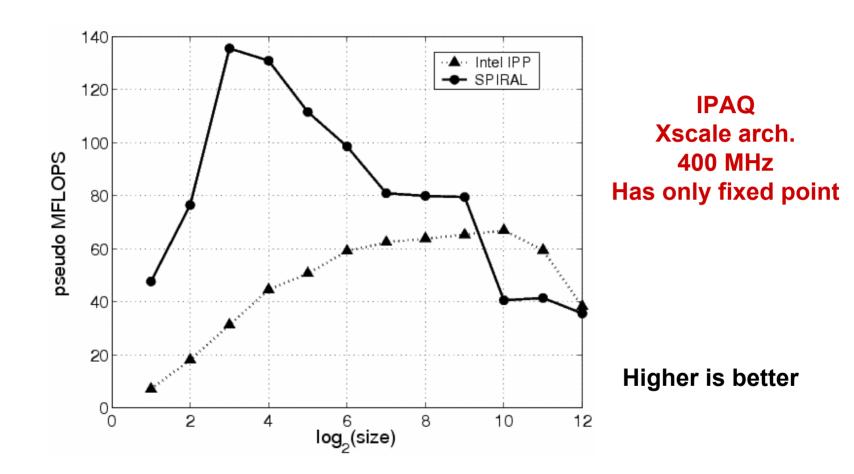
Benchmark: DFT, Level 2 Sizes



Benchmark: DFT, Level 3 Sizes



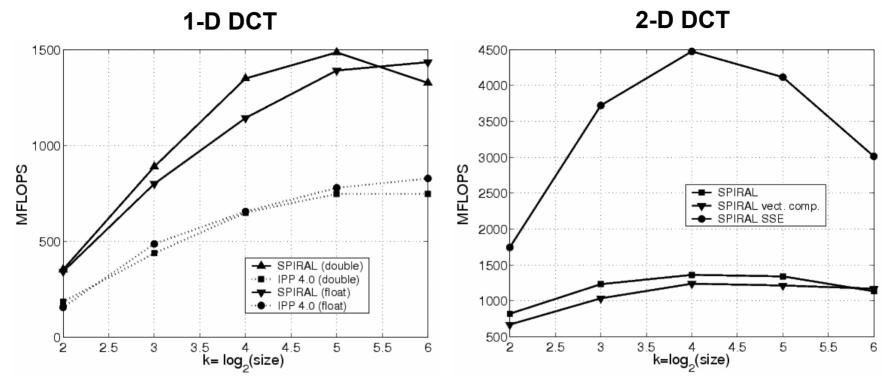
Benchmark: Fixed Point DFT, IPAQ



Intel spent less effort?

Benchmark: DCT

P4, 3.2 GHz, icc 8.0

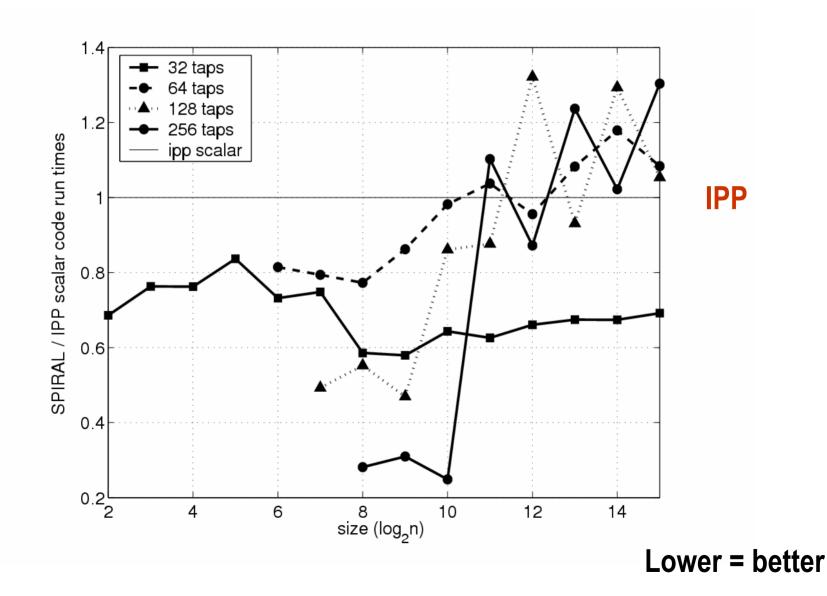


Scalar code

Scalar vs. SSE code

- This is not the latest IPP
- Spiral gains a factor of 2 to vendor library
- Another factor of 3 with 2D and vector instructions

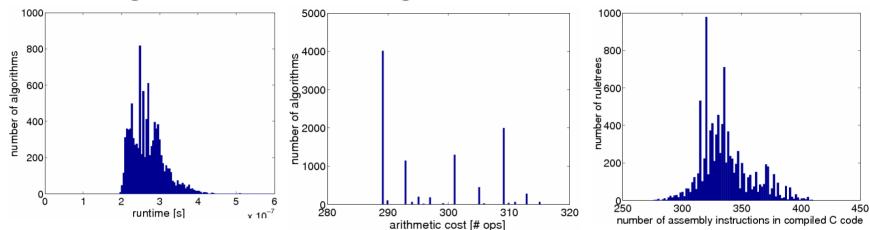
Benchmark: Filter (Relative to IPP)



Instructive Experiments

Performance Spread: DCT, size 32 Histograms, 10,000 algorithms

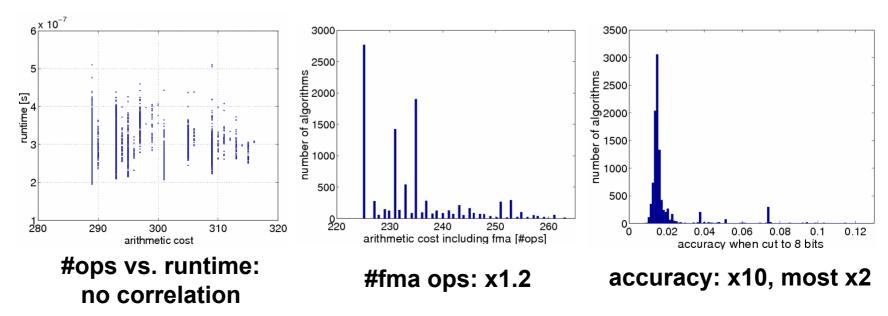
Carnegie Mellon



runtime: x2

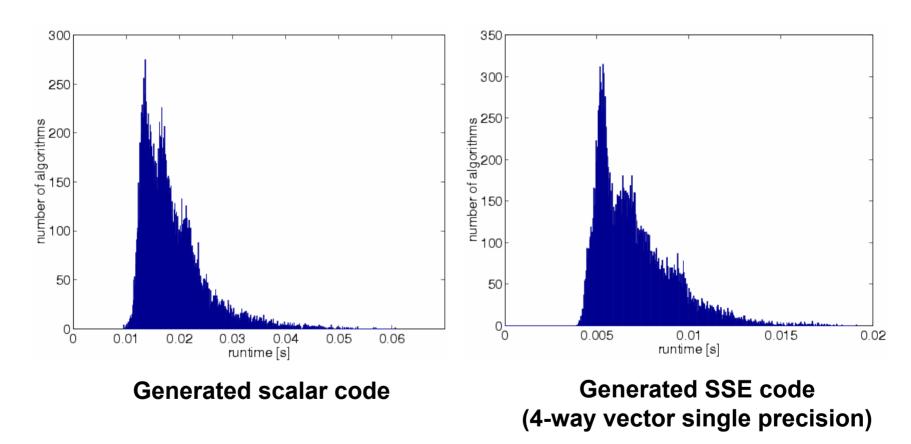
#ops: x1.08

#assembly instr: x1.5



Performance Spread: DFT 2^16 Histograms, 20,000 Algorithms

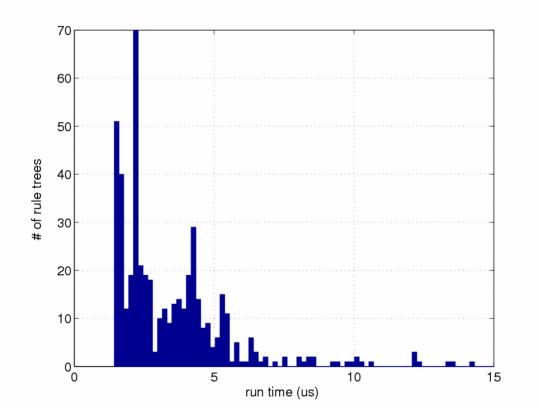
P4, 3.2 GHz, icc 8.0



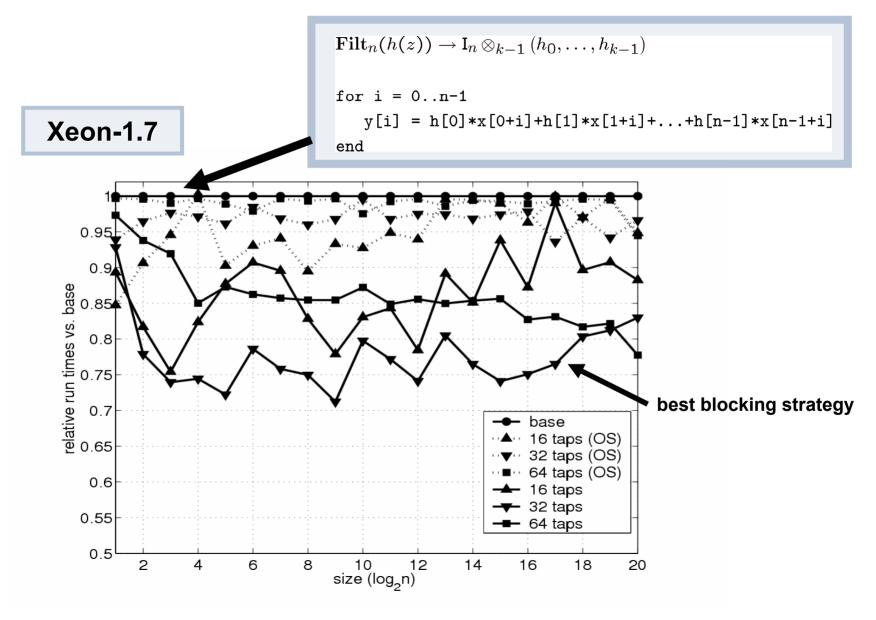
- Generality of vectorization (all algorithms improve)
- Efficiency of vectorization (x 2.5 gain)

Performance Spread: Filter(128, 16)

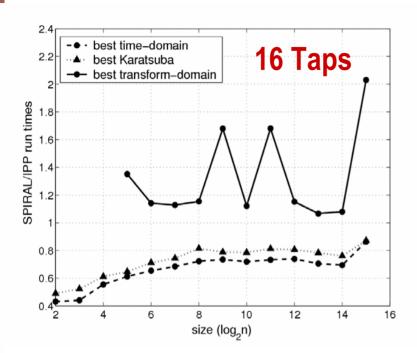
Pentium 4 – 3.2



Filter: Time Domain Methods

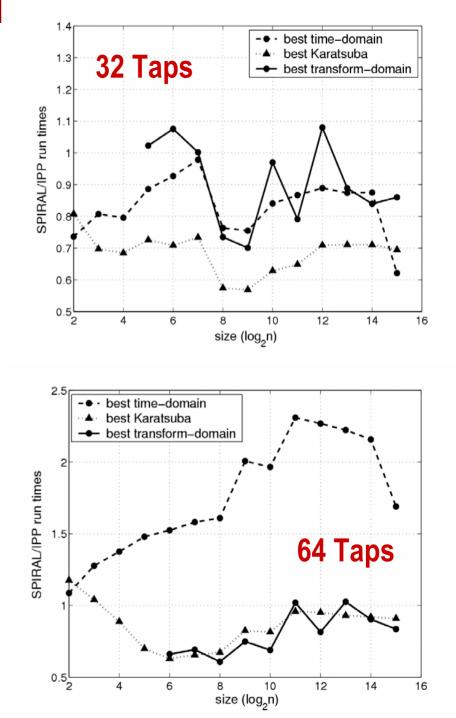


Filter: All Methods

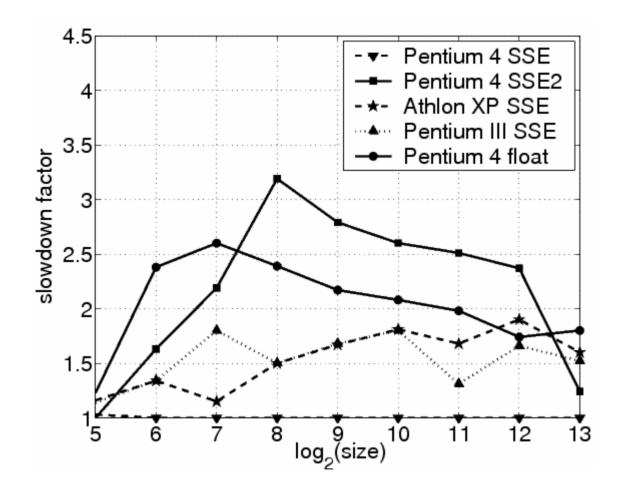


Athlon XP 1.73

- 16: Time domain wins
- 32: Karatsuba wins
- · 64: Karatsuba/DFT ~equal



Platform Dependency: DFT



50% Loss by porting from PIII to P4

Platform Dependency: Filter

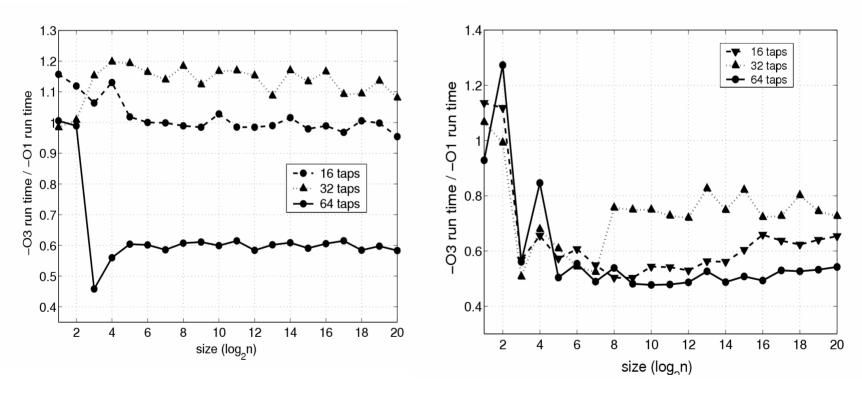
| | 16-tap | 32-tap | 64-tap | 128-tap |
|-------------------------------|------------------------|-----------|--------------------|---------|
| Pentium 4 3.0GHz Northwood | Blocking | Karatsuba | RDFT | RDFT |
| Pentium 4 3.6GHz Prescott | Blocking | Karatsuba | Karatsuba | RDFT |
| Macintosh | Karatsuba | Karatsuba | RDFT | RDFT |
| Xeon 1.7 GHz | Blocking | Blocking | Blocking | RDFT |
| Athlon 1.73GHz | Karatsuba/ Blocking | Karatsuba | Karatsuba/ RDFT | RDFT |

Compileroptions: Filter

Macintosh - GNU C 3.3 (Apple)

Blocking/nesting

+ Karatsuba

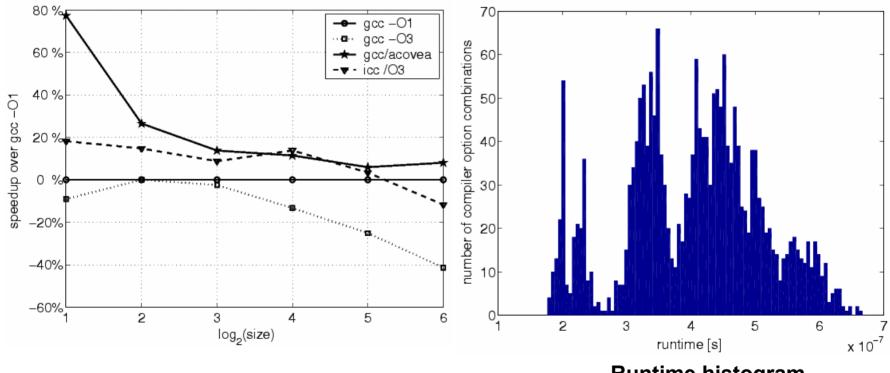


gcc {-01/-03} -fomit-frame-pointer -std=c99 -fast -mcpu=7450

Compileroptions DCT

P4, 3.2 GHz, gcc

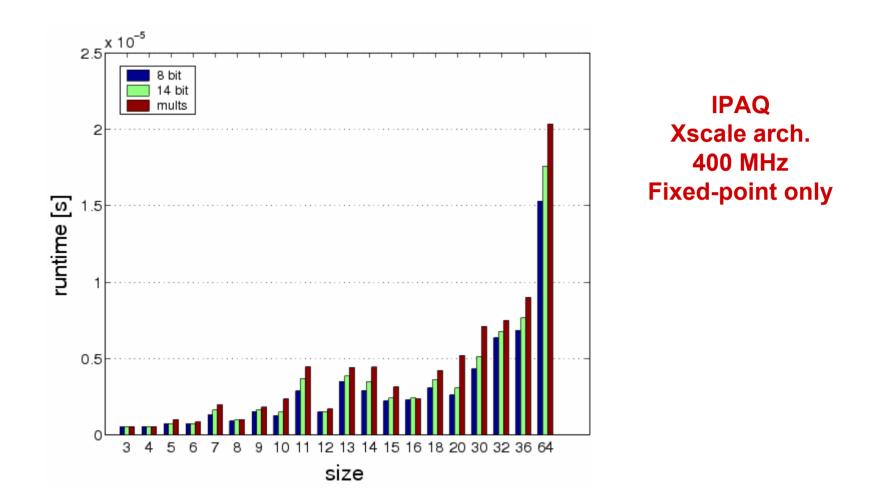
ACOVEA: Evolutionary search for best compiler flags (gcc has ~500)



Runtime histogram Random compiler flags incl. -O1 –march=pentium4

10% improvement of best Spiral generated code

Multiplierless DFT, IPAQ



- fixed-point constant multiplications replaced by adds and shifts
- trade-off runtime and precision

Summary

Code generation and tuning as optimization problem over the algorithm and implementation space

Exploit the structure of the domain to solve it

Declarative framework for computer representation of the domainknowledge

> *Enables algorithm generation and optimization (teaches the system the math; does not become obsolete?)*

- Compiler to translate math description into code
- Search and learning to explore implementation space

Closes the feedback loop gives the system "intelligence"

Extensible, versatile

Every step in the code generation is exposed

www.spiral.net