

# How to Write Fast Numerical Code

Spring 2015

*Lecture:* Benchmarking, Compiler Limitations

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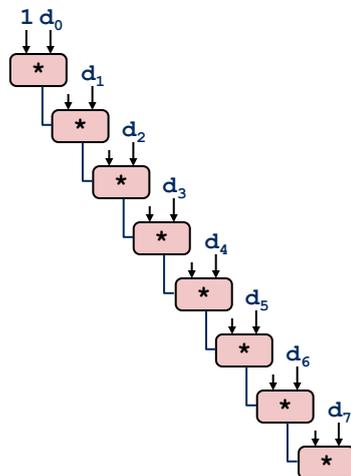
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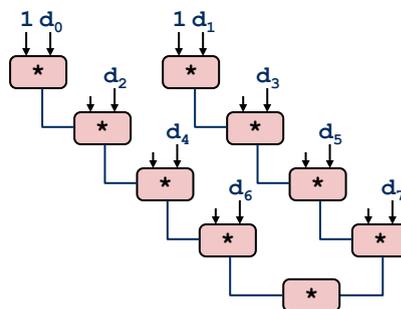
Eidgenössische Technische Hochschule Zürich  
Swiss Federal Institute of Technology Zurich

## Last Time: ILP

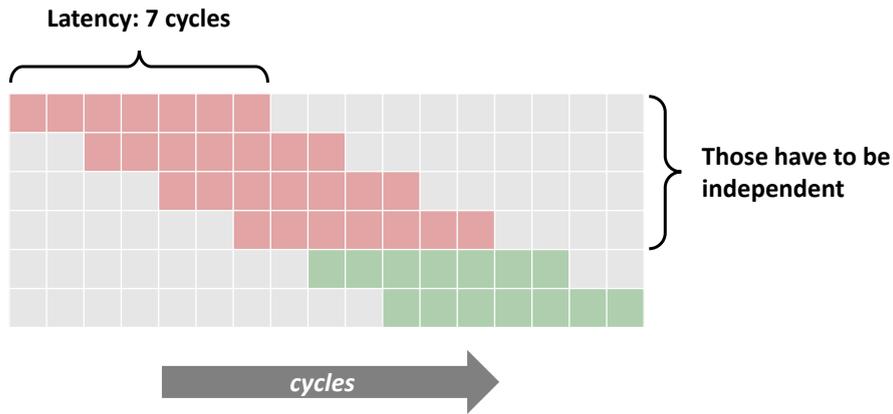
- Latency/throughput (Pentium 4 fp mult: 7/2)



*Twice as fast*



## Last Time: How Many Accumulators?



Based on this insight:  $K = \text{\#accumulators} = \text{ceil}(\text{latency}/\text{cycles per issue})$

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## Compiler Limitations

```
void combine4(vec_ptr v, data_t *dest)
{
    int i;
    int length = vec_length(v);
    data_t *d = get_vec_start(v);
    data_t t = IDENT;
    for (i = 0; i < length; i++)
        t = t OP d[i];
    *dest = t;
}
```



```
void unroll2_sa(vec_ptr v, data_t *dest)
{
    int length = vec_length(v);
    int limit = length-1;
    data_t *d = get_vec_start(v);
    data_t x0 = IDENT;
    data_t x1 = IDENT;
    int i;
    /* Combine 2 elements at a time */
    for (i = 0; i < limit; i+=2)
        x0 = x0 OP d[i];
        x1 = x1 OP d[i+1];
    /* Finish any remaining elements */
    for (; i < length; i++)
        x0 = x0 OP d[i];
    *dest = x0 OP x1;
}
```

- Associativity law does not hold for floats: illegal transformation
- No good way of handling choices (e.g., number of accumulators)
- *More examples of limitations today*

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# Today

- **Measuring performance & benchmarking**

*Section 3.2 in the tutorial*

<http://spiral.ece.cmu.edu:8080/pub-spiral/abstract.jsp?id=100>

- **Optimizing compilers and optimization blockers**

- Overview
- Code motion
- Strength reduction
- Sharing of common subexpressions
- Removing unnecessary procedure calls
- Optimization blocker: Procedure calls
- Optimization blocker: Memory aliasing
- Summary

*Chapter 5 in **Computer Systems: A Programmer's Perspective**, 2<sup>nd</sup> edition,  
Randal E. Bryant and David R. O'Hallaron, Addison Wesley 2010*

*Part of these slides are adapted from the course associated with this book*

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# Benchmarking

- **First: Validate/test your code!**

- **Measure runtime (in [s] or [cycles]) for a set of relevant input sizes**

- seconds: actual runtime
- cycles: abstracts from CPU frequency

- **Usually: Compute and show performance (in [flop/s] or [flop/cycle])**

- **Careful: Better performance ≠ better runtime (why?)**

- Op count could differ
- Never show in one plot performance of two algorithms with substantially different op count

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## How to Measure Runtime?

- **C clock()**
  - process specific, low resolution, very portable
- **gettimeofday**
  - measures wall clock time, higher resolution, somewhat portable
- **Performance counter (e.g., TSC on Intel)**
  - measures cycles (i.e., also wall clock time), highest resolution, not portable
- **Careful:**
  - measure only what you want to measure
  - ensure proper machine state  
(e.g., cold or warm cache = input data is or is not in cache)
  - measure enough repetitions
  - check how reproducible; if not reproducible: fix it
- ***Getting proper measurements is not easy at all!***

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## Problems with Timing

- **Too few iterations: inaccurate non-reproducible timing**
- **Too many iterations: system events interfere**
- **Machine is under load: produces side effects**
- **Multiple timings performed on the same machine**
- **Bad data alignment of input/output vectors:**
  - align to multiples of cache line (on Core: address is divisible by 64)
  - sometimes aligning to page boundaries (address divisible by 4096) makes sense
- **Machine was not rebooted for a long time: state of operating system causes problems**
- **Computation is input data dependent: choose representative input data**
- **Computation is inplace and data grows until an exception is triggered (computation is done with NaNs)**
- **You work on a computer that has dynamic frequency scaling (e.g., turbo boost)**
- ***Always check whether timings make sense, are reproducible***

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# Benchmarks in Writing

- **Specify experimental setup**

- platform
- compiler and version
- compiler flags used

- **Plot: Very readable**

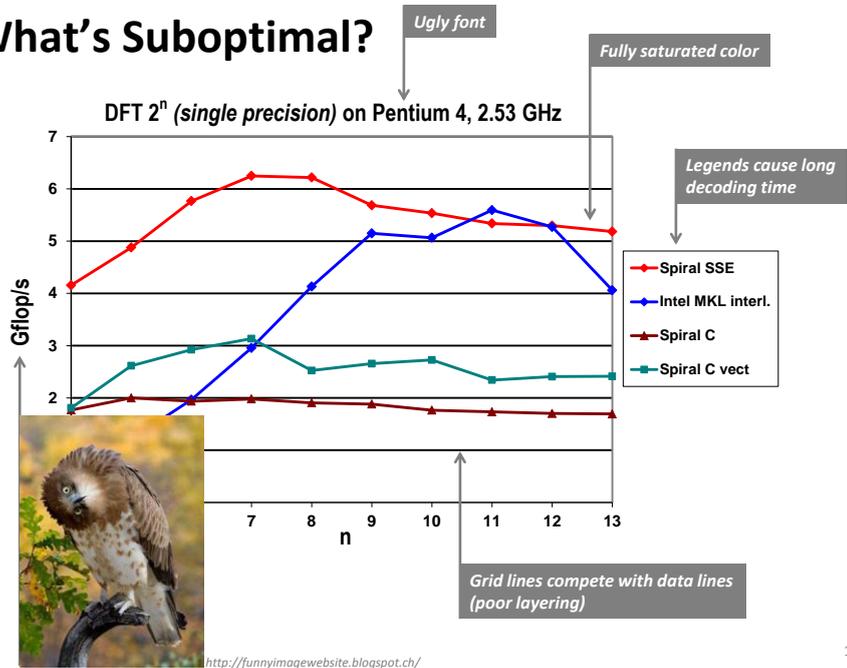
- Title, x-label, y-label should be there
- Fonts large enough
- Enough contrast (e.g., no yellow on white please)
- Proper number format

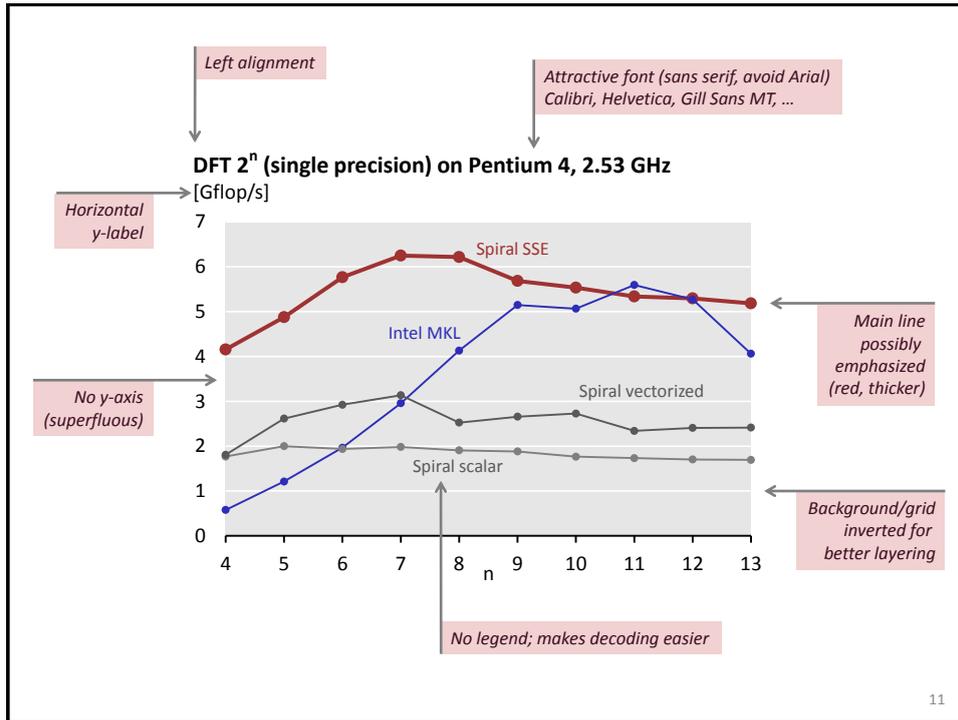
*No:* 13.254687; *yes:* 13.25

*No:* 2.0345e-05 s; *yes:* 20.3  $\mu$ s

*No:* 100000 B; *maybe:* 100,000 B; *yes:* 100 KB

# What's Suboptimal?





## Today

- **Measuring performance & benchmarking**
- **Optimizing compilers and optimization blockers**
  - Overview
  - Removing unnecessary procedure calls
  - Code motion
  - Strength reduction
  - Sharing of common subexpressions
  - Optimization blocker: Procedure calls
  - Optimization blocker: Memory aliasing
  - Summary

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# Optimizing Compilers



- **Always use optimization flags:**
  - gcc: *default is no optimization* (-O0)!
  - icc: some optimization is turned on
- **Good choices for gcc/icc: -O2, -O3, -march=xxx, -mSSE3, -m64**
  - Read in manual what they do
  - Try to understand the differences
- **Try different flags and maybe different compilers**

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## Example (On Core 2 Duo)

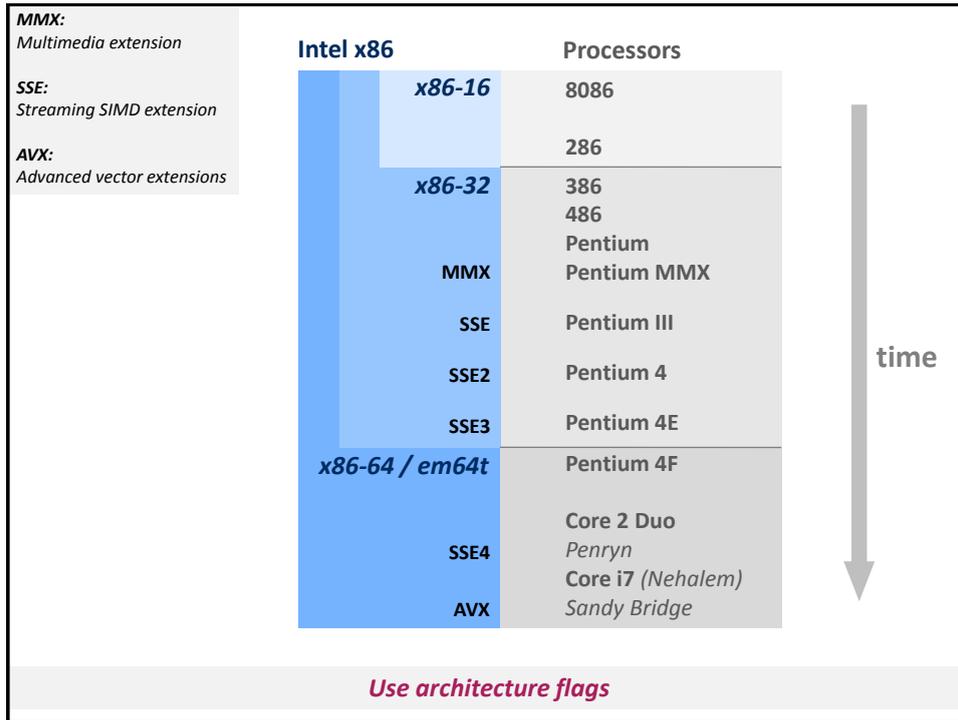
```
double a[4][4];
double b[4][4];
double c[4][4];

/* Multiply 4 x 4 matrices c = a*b + c */
void mmm(double *a, double *b, double *c) {
    int i, j, k;

    for (i = 0; i < 4; i++)
        for (j = 0; j < 4; j++)
            for (k = 0; k < 4; k++)
                c[i*4+j] += a[i*4 + k]*b[k*4 + j];
}
```

- **Compiled without flags:**  
~1300 cycles
- **Compiled with -O3 -m64 -march=... -fno-tree-vectorize** ← *Prevents use of SSE*  
~150 cycles

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## Optimizing Compilers

- Compilers are *good* at: mapping program to machine
  - register allocation
  - code selection and ordering (instruction scheduling)
  - dead code elimination
  - eliminating minor inefficiencies
- Compilers are *not good* at: algorithmic restructuring
  - for example to increase ILP, locality, etc.
  - cannot deal with choices
- Compilers are *not good* at: overcoming “optimization blockers”
  - potential memory aliasing
  - potential procedure side-effects

## Limitations of Optimizing Compilers

- *If in doubt, the compiler is conservative*
- **Operate under fundamental constraints**
  - Must not change program behavior under any possible condition
  - Often prevents it from making optimizations when would only affect behavior under pathological conditions
- **Most analysis is performed only within procedures**
  - Whole-program analysis is too expensive in most cases
- **Most analysis is based only on *static* information**
  - Compiler has difficulty anticipating run-time inputs
  - Not good at evaluating or dealing with choices

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## Organization

- Instruction level parallelism (ILP): an example
- **Optimizing compilers and optimization blockers**
  - Overview
  - *Code motion*
  - *Strength reduction*
  - *Sharing of common subexpressions*
  - Removing unnecessary procedure calls
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## Code Motion

- Reduce frequency with which computation is performed
  - If it will always produce same result
  - Especially moving code out of loop (loop-invariant code motion)
- Sometimes also called precomputation

```
void set_row(double *a, double *b,
            int i, int n)
{
    int j;
    for (j = 0; j < n; j++)
        a[n*i+j] = b[j];
}
```



```
int j;
int ni = n*i;
for (j = 0; j < n; j++)
    a[ni+j] = b[j];
```

- Compiler is likely to do

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## Strength Reduction

- Replace costly operation with simpler one
- Example: Shift/add instead of multiply or divide  $16*x \rightarrow x \ll 4$ 
  - Utility machine dependent
- Example: Recognize sequence of products

```
for (i = 0; i < n; i++)
    for (j = 0; j < n; j++)
        a[n*i + j] = b[j];
```



```
int ni = 0;
for (i = 0; i < n; i++) {
    for (j = 0; j < n; j++)
        a[ni + j] = b[j];
    ni += n;
}
```

- Compiler is likely to do

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## Share Common Subexpressions

- Reuse portions of expressions
- Compilers often not very sophisticated in exploiting arithmetic properties

**3 mults:  $i*n$ ,  $(i-1)*n$ ,  $(i+1)*n$**

```
/* Sum neighbors of i,j */
up   = val[(i-1)*n + j ];
down = val[(i+1)*n + j ];
left = val[i*n     + j-1];
right = val[i*n     + j+1];
sum  = up + down + left + right;
```

**1 mult:  $i*n$**

```
int inj = i*n + j;
up      = val[inj - n];
down    = val[inj + n];
left    = val[inj - 1];
right   = val[inj + 1];
sum     = up + down + left + right;
```

- In simple cases compiler is likely to do

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*Compiler is likely to do*

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## Example: Data Type for Vectors

```
/* data structure for vectors */
typedef struct{
    int len;
    double *data;
} vec;
```



```
/* retrieve vector element and store at val */
int get_vec_element(vec *v, int idx, double *val)
{
    if (idx < 0 || idx >= v->len)
        return 0;
    *val = v->data[idx];
    return 1;
}
```

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## Example: Summing Vector Elements

```
/* retrieve vector element and store at val */
int get_vec_element(vec *v, int idx, double *val)
{
    if (idx < 0 || idx >= v->len)
        return 0;
    *val = v->data[idx];
    return 1;
}
```

```
/* sum elements of vector */
double sum_elements(vec *v, double *res)
{
    int i;
    n = vec_length(v);
    *res = 0.0;
    double t;

    for (i = 0; i < n; i++) {
        get_vec_element(v, i, &t);
        *res += t;
    }
    return res;
}
```

### Overhead for every fp +:

- One fct call
- One <
- One >=
- One ||
- One memory variable access

*Potential big performance loss*

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## Removing Procedure Call

```
/* sum elements of vector */
double sum_elements(vec *v, double *res)
{
    int i;
    n = vec_length(v);
    *res = 0.0;
    double t;

    for (i = 0; i < n; i++) {
        get_vec_element(v, i, &t);
        *res += t;
    }
    return res;
}
```

```
/* sum elements of vector */
double sum_elements_opt(vec *v, double *res)
{
    int i;
    n = vec_length(v);
    *res = 0.0;
    double *data = get_vec_start(v);

    for (i = 0; i < n; i++)
        *res += data[i];
    return res;
}
```

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## Removing Procedure Calls

- Procedure calls can be very expensive
- Bound checking can be very expensive
- Abstract data types can easily lead to inefficiencies
  - Usually avoided in superfast numerical library functions
- *Watch your innermost loop!*
- *Get a feel for overhead versus actual computation being performed*

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## Further Inspection of the Example

```
vector.c // vector data type      Intel Xeon E3-1285L v3 (Haswell)
sum.c    // sum                    CC=gcc -w -O3 -std=c99 -march=core-avx2
sum_opt.c // optimized sum        Intel Atom D2550
main.c   // timing                 CC=gcc -w -std=c99 -O3 -march=atom
```

```
$(CC) -c -o vector.o vector.c
$(CC) -c -o sum.o sum.c
$(CC) -c -o main.o main.c
$(CC) -o vector vector.o sum.o main.o
```

**Xeon:** 7.2 cycles/add  
**Atom:** 28 cycles/add

```
$(CC) -c -o vector.o vector.c
$(CC) -c -o sum_opt.o sum_opt.c
$(CC) -c -o main.o main.c
$(CC) -o vector vector.o sum_opt.o main.o
```

**Xeon:** 2.4 cycles/add  
**Atom:** 6 cycles/add

```
$(CC) -c -o vector.o vector.c sum.c
$(CC) -c -o main.o main.c
$(CC) -o vector vector.o main.o
```

**Xeon:** 2.4 cycles/add  
**Atom:** 6 cycles/add

*What's happening here?*

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## Function Inlining

- Compilers may be able to do function inlining
  - Replace function call with body of function
  - Usually requires that source code is compiled together

```
/* retrieve vector element and store at val */
int get_vec_element(vec *v, int idx, double *val)
{
    if (idx < 0 || idx >= v->len)
        return 0;
    *val = v->data[idx];
    return 1;
}
```

insert

```
/* sum elements of vector */
double sum_elements(vec *v, double *res)
{
    int i;
    n = vec_length(v);
    *res = 0.0;
    double t;
    for (i = 0; i < n; i++) {
        get_vec_element(v, i, &t);
        *res += t;
    }
    return res;
}
```

- Enables other optimizations
- Problem:** performance libraries distributed as binary

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# Organization

- Instruction level parallelism (ILP): an example
- Optimizing compilers and optimization blockers
  - Overview
  - Code motion *Compiler is likely to do*
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  - **Optimization blocker: Procedure calls**
  - Optimization blocker: Memory aliasing
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# Optimization Blocker #1: Procedure Calls

- Procedure to convert string to lower case

```
void lower(char *s)
{
    int i;
    for (i = 0; i < strlen(s); i++)
        if (s[i] >= 'A' && s[i] <= 'Z')
            s[i] -= ('A' - 'a');
}
```

*O(n<sup>2</sup>) instead of O(n)*

```
/* My version of strlen */
size_t strlen(const char *s)
{
    size_t length = 0;
    while (*s != '\0') {
        s++;
        length++;
    }
    return length;
}
```

*O(n)*

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## Improving Performance

```
void lower(char *s)
{
    int i;
    for (i = 0; i < strlen(s); i++)
        if (s[i] >= 'A' && s[i] <= 'Z')
            s[i] -= ('A' - 'a');
}
```

```
void lower(char *s)
{
    int i;
    int len = strlen(s);
    for (i = 0; i < len; i++)
        if (s[i] >= 'A' && s[i] <= 'Z')
            s[i] -= ('A' - 'a');
}
```

- Move call to `strlen` outside of loop
- Form of code motion/precomputation

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## Optimization Blocker: Procedure Calls

- Why couldn't compiler move `strlen` out of inner loop?
  - Procedure may have side effects
- *Compiler usually treats procedure call as a black box that cannot be analyzed*
  - Consequence: conservative in optimizations
- In this case the compiler may actually do it if `strlen` is recognized as built-in function whose properties are known

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# Organization

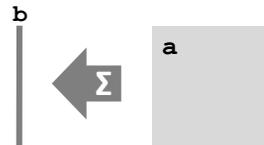
- Instruction level parallelism (ILP): an example
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  - Removing unnecessary procedure calls *Compiler is likely to do*
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# Optimization Blocker: Memory Aliasing

```
/* Sums rows of n x n matrix a
   and stores in vector b */
void sum_rows1(double *a, double *b, int n) {
    int i, j;

    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}
```



- Code updates `b[i]` (= memory access) on every iteration

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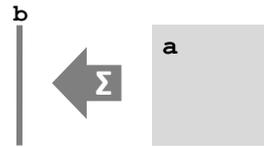
# Optimization Blocker: Memory Aliasing

```

/* Sums rows of n x n matrix a
and stores in vector b */
void sum_rows1(double *a, double *b, int n) {
    int i, j;

    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}

```



```

/* Sums rows of n x n matrix a
and stores in vector b */
void sum_rows2(double *a, double *b, int n) {
    int i, j;

    for (i = 0; i < n; i++) {
        double val = 0;
        for (j = 0; j < n; j++)
            val += a[i*n + j];
        b[i] = val;
    }
}

```

Does compiler optimize this?

**No!**  
**Why?**

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# Reason: Possible Memory Aliasing

- If memory is accessed, compiler assumes the possibility of side effects
- Example:

```

/* Sums rows of n x n matrix a
and stores in vector b */
void sum_rows1(double *a, double *b, int n) {
    int i, j;

    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}

```

```

double A[9] =
{ 0, 1, 2,
  4, 8, 16,
  32, 64, 128};

double B[3] = A+3;

sum_rows1(A, B, 3);

```

Value of B:

init: [4, 8, 16]

i = 0: [3, 8, 16]

i = 1: [3, 22, 16]

i = 2: [3, 22, 224]

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## Removing Aliasing

```
/* Sums rows of n x n matrix a
   and stores in vector b */
void sum_rows2(double *a, double *b, int n) {
    int i, j;

    for (i = 0; i < n; i++) {
        double val = 0;
        for (j = 0; j < n; j++)
            val += a[i*n + j];
        b[i] = val;
    }
}
```

### ■ Scalar replacement:

- Copy array elements *that are reused* into temporary variables
- Perform computation on those variables
- Enables register allocation and instruction scheduling
- Assumes no memory aliasing (otherwise incorrect)

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## Optimization Blocker: Memory Aliasing

### ■ *Memory aliasing:*

Two different memory references write to the same location

### ■ Easy to have happen in C

- Since allowed to do address arithmetic
- Direct access to storage structures

### ■ Hard to analyze = compiler cannot figure it out

- Hence is conservative

### ■ Solution: *Scalar replacement* in innermost loop

- Copy memory variables that are reused into local variables
- Basic scheme:

**Load:**  $t1 = a[i]$ ,  $t2 = b[i+1]$ , ...

**Compute:**  $t4 = t1 * t2$ ; ...

**Store:**  $a[i] = t12$ ,  $b[i+1] = t7$ , ...

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## Example: MMM

Which array elements are reused? *All of them! But how to take advantage?*

```
void mmm(double const * A, double const * B, double * C, size_t N) {  
    for( size_t k6 = 0; k6 < N; k6++ )  
        for( size_t i5 = 0; i5 < N; i5++ )  
            for( size_t j7 = 0; j7 < N; j7++ )  
                C[N*i5 + j7] = C[N*i5 + j7] + A[N*i5 + k6] * B[j7 + N*k6]; }  
}
```

*tile each loop (= blocking MMM)*

```
void mmm(double const * A, double const * B, double * C, size_t N) {  
    for( size_t i21 = 0; i21 < N; i21+=2 )  
        for( size_t j23 = 0; j23 < N; j23+=2 )  
            for( size_t k22 = 0; k22 < N; k22+=2 )  
                for( size_t kk25 = 0; kk25 < 2; kk25++ )  
                    for( size_t ii24 = 0; ii24 < 2; ii24++ )  
                        for( size_t jj26 = 0; jj26 < 2; jj26++ )  
                            C[N*i21 + N*ii24 + j23 + jj26] = C[N*i21 + N*ii24 + j23 + jj26] +  
                                A[N*i21 + N*ii24 + k22 + kk25] * B[j23 + jj26 + N*k22 + N*kk25]; }  
}
```

*unroll inner three loops*

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Now the reuse becomes apparent  
(every elements used twice)

*unroll inner three loops*

```
void mmm(double const * A, double const * B, double * C, size_t N) {  
    for( size_t i21 = 0; i21 < N; i21+=2 )  
        for( size_t j23 = 0; j23 < N; j23+=2 )  
            for( size_t k22 = 0; k22 < N; k22+=2 ) {  
                C[N*i21 + j23] = C[N*i21 + j23] + A[N*i21 + k22] * B[j23 + N*k22];  
                C[N*i21 + j23 + 1] = C[N*i21 + j23 + 1] + A[N*i21 + k22] * B[j23 + N*k22 + 1];  
                C[N*i21 + N + j23] = C[N*i21 + N + j23] + A[N*i21 + N + k22] * B[j23 + N*k22];  
                C[N*i21 + N + j23 + 1] = C[N*i21 + N + j23 + 1] + A[N*i21 + N + k22] * B[j23 + N*k22 + 1];  
                C[N*i21 + j23] = C[N*i21 + j23] + A[N*i21 + k22 + 1] * B[j23 + N*k22 + N];  
                C[N*i21 + j23 + 1] = C[N*i21 + j23 + 1] + A[N*i21 + k22 + 1] * B[j23 + N*k22 + N + 1];  
                C[N*i21 + N + j23] = C[N*i21 + N + j23] + A[N*i21 + N + k22 + 1] * B[j23 + N*k22 + N];  
                C[N*i21 + N + j23 + 1] =  
                    C[N*i21 + N + j23 + 1] + A[N*i21 + N + k22 + 1] * B[j23 + N*k22 + N + 1];  
            }  
}
```

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Now the reuse becomes apparent  
(every elements used twice)

*unroll inner three loops*

```
void mmm(double const * A, double const * B, double * C, size_t N) {
    for( size_t i21 = 0; i21 < N; i21+=2 )
        for( size_t j23 = 0; j23 < N; j23+=2 )
            for( size_t k22 = 0; k22 < N; k22+=2 ) {
                C[N*i21 + j23] = C[N*i21 + j23] + A[N*i21 + k22] * B[j23 + N*k22];
                C[N*i21 + j23 + 1] = C[N*i21 + j23 + 1] + A[N*i21 + k22] * B[j23 + N*k22 + 1];
                C[N*i21 + N + j23] = C[N*i21 + N + j23] + A[N*i21 + N + k22] * B[j23 + N*k22];
                C[N*i21 + N + j23 + 1] = C[N*i21 + N + j23 + 1] + A[N*i21 + N + k22] * B[j23 + N*k22 + 1];
                C[N*i21 + j23] = C[N*i21 + j23] + A[N*i21 + k22 + 1] * B[j23 + N*k22 + N];
                C[N*i21 + j23 + 1] = C[N*i21 + j23 + 1] + A[N*i21 + k22 + 1] * B[j23 + N*k22 + N + 1];
                C[N*i21 + N + j23] = C[N*i21 + N + j23] + A[N*i21 + N + k22 + 1] * B[j23 + N*k22 + N];
                C[N*i21 + N + j23 + 1] =
                    C[N*i21 + N + j23 + 1] + A[N*i21 + N + k22 + 1] * B[j23 + N*k22 + N + 1];
            }
    }
}
```

*scalar replacement*

```
void mmm(double const * A, double const * B, double * C, size_t N) {
    for( size_t i21 = 0; i21 < N; i21+=2 )
        for( size_t j23 = 0; j23 < N; j23+=2 )
            for( size_t k22 = 0; k22 < N; k22+=2 ) {

                double t0_0, t0_1, t0_2, t0_3, t0_4, t0_5, t0_6, t0_7, t0_8, t0_9, t0_10, t0_11, t0_12;

                t0_7 = A[N*i21 + k22];
                t0_6 = A[N*i21 + k22 + 1];
                t0_5 = A[N*i21 + N + k22];
                t0_4 = A[N*i21 + N + k22 + 1];
                t0_3 = B[j23 + N*k22];
                t0_2 = B[j23 + N*k22 + 1];
                t0_1 = B[j23 + N*k22 + N];
                t0_0 = B[j23 + N*k22 + N + 1];
                t0_8 = C[N*i21 + j23];
                t0_9 = C[N*i21 + j23 + 1];
                t0_10 = C[N*i21 + N + j23];
                t0_11 = C[N*i21 + N + j23 + 1];
                t0_12 = t0_7 * t0_3;
                t0_8 = t0_8 + t0_12;
                t0_12 = t0_7 * t0_2;
                t0_9 = t0_9 + t0_12;
                t0_12 = t0_5 * t0_3;
                t0_10 = t0_10 + t0_12;
                t0_12 = t0_5 * t0_2;
                t0_11 = t0_11 + t0_12;
                t0_12 = t0_6 * t0_1;
                t0_8 = t0_8 + t0_12;
                t0_12 = t0_6 * t0_0;
                t0_9 = t0_9 + t0_12;
                t0_12 = t0_4 * t0_1;
                t0_10 = t0_10 + t0_12;
                t0_12 = t0_4 * t0_0;
                t0_11 = t0_11 + t0_12;
                C[N*i21 + j23] = t0_8;
                C[N*i21 + j23 + 1] = t0_9;
                C[N*i21 + N + j23] = t0_10;
                C[N*i21 + N + j23 + 1] = t0_11;
            }
    }
}
```

*load*

*compute*

*store*

## Effect on Runtime?

Intel Core i7-2600 (Sandy Bridge)  
compiler: icc 12.1  
flags: -O3 -no-vec -no-ipo -no-ip

	N = 4	N = 100
Triple loop	202	2.3M

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## Effect on Runtime?

Intel Core i7-2600 (Sandy Bridge)  
compiler: icc 12.1  
flags: -O3 -no-vec -no-ipo -no-ip

	N = 4	N = 100
Triple loop	202	2.3M
Six-fold loop	144	2.3M
+ Inner three unrolled	166	2.4M
+ scalar replacement	106	1.6M

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## Can Compiler Remove Aliasing?

```
for (i = 0; i < n; i++)  
    a[i] = a[i] + b[i];
```

Potential aliasing: Can compiler do something about it?

Compiler can insert runtime check:

```
if (a + n < b || b + n < a)  
    /* further optimizations may be possible now */  
    ...  
else  
    /* aliased case */  
    ...
```

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## Removing Aliasing With Compiler

- Globally with compiler flag:
  - -fno-alias, /Oa
  - -fargument-noalias, /Qalias-args- (function arguments only)

- For one loop: pragma

```
void add(float *a, float *b, int n) {  
    #pragma ivdep  
    for (i = 0; i < n; i++)  
        a[i] = a[i] + b[i];  
}
```

- For specific arrays: restrict (needs compiler flag -restrict, /Qrestrict)

```
void add(float *restrict a, float *restrict b, int n) {  
    for (i = 0; i < n; i++)  
        a[i] = a[i] + b[i];  
}
```

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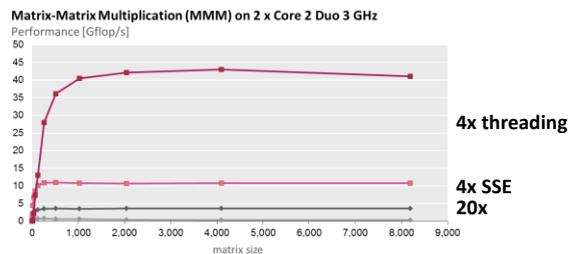
# Organization

- Instruction level parallelism (ILP): an example
- Optimizing compilers and optimization blockers
  - Overview
  - Removing unnecessary procedure calls *Compiler is likely to do*
  - Code motion
  - Strength reduction
  - Sharing of common subexpressions
  - Removing unnecessary procedure calls
  - Optimization blocker: Procedure calls
  - Optimization blocker: Memory aliasing
  - *Summary*

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# Summary

- *One can easily loose 10x, 100x in runtime or even more*



- What matters besides operation count:
  - Coding style (unnecessary procedure calls, unrolling + ..., reordering, ...)
  - Algorithm structure (instruction level parallelism, locality, ...)
  - Data representation (complicated structs or simple arrays)

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## Summary: Optimize at Multiple Levels

- **Algorithm:**
  - Evaluate different algorithm choices
  - Restructuring may be needed (ILP, locality)
- **Data representations:**
  - Careful with overhead of complicated data types
  - Best are arrays
- **Procedures:**
  - Careful with overhead
  - They are black boxes for the compiler
- **Loops:**
  - Often need to be restructured (ILP, locality)
  - Unrolling often necessary *to enable other optimizations*
  - Watch the innermost loop bodies

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## Numerical Functions

- **Use arrays (simple data structure) if possible**
- **Unroll to some extent**
  - To make ILP explicit
  - To enable scalar replacement and hence register allocation for variables that are reused

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