Your Optimizing Compiler is not Optimizing Enough. To hell with Multiple Recursions!

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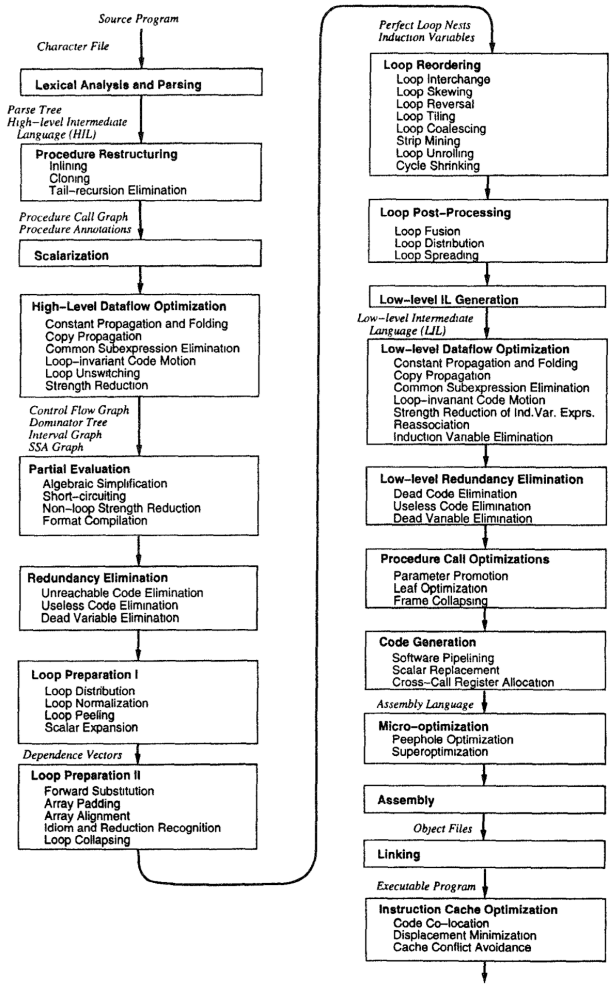
Compilers as Musical Compositions

Compilers are frequently perceived as intricate musical compositions—like the unfinished *J. S. Bach's Art of Fugue*—where mathematical precision and logical interplay guide each part.

Every module enters in perfect timing, weaving together a structure that only the keenest ears can fully grasp.

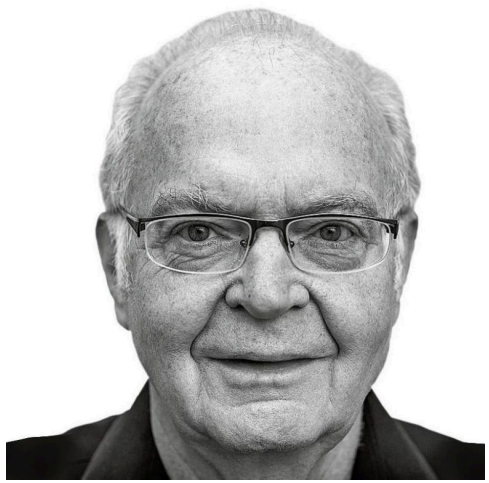


Bacon *et al.*, CSUR 1994 [1]

Premature Optimizations

Donald E. Knuth warned in 1974 about the dangers of **premature optimization** in programming [2]:

We should forget about small efficiencies, say about 97% of the time: premature optimization is the root of all evil. Yet we should not pass up our opportunities in that critical 3%.



In the absence of either empirically measured or theoretically justified performance issues, programmers should **avoid** making optimizations based **solely** on assumptions about potential performance gains.

Optimizing Compilers

Compilers use information collected during *analysis passes* to guide *transformations*.

Compiler optimizations are such transformations (say meaning-preserving mappings [164]) applied to the input code to improve certain aspects—such as performance, resource utilization, and power consumption [2, 159]—without altering its observable behavior [80].

In accordance with the literature [6, 214], such compilers are referred to as **optimizing compilers**.

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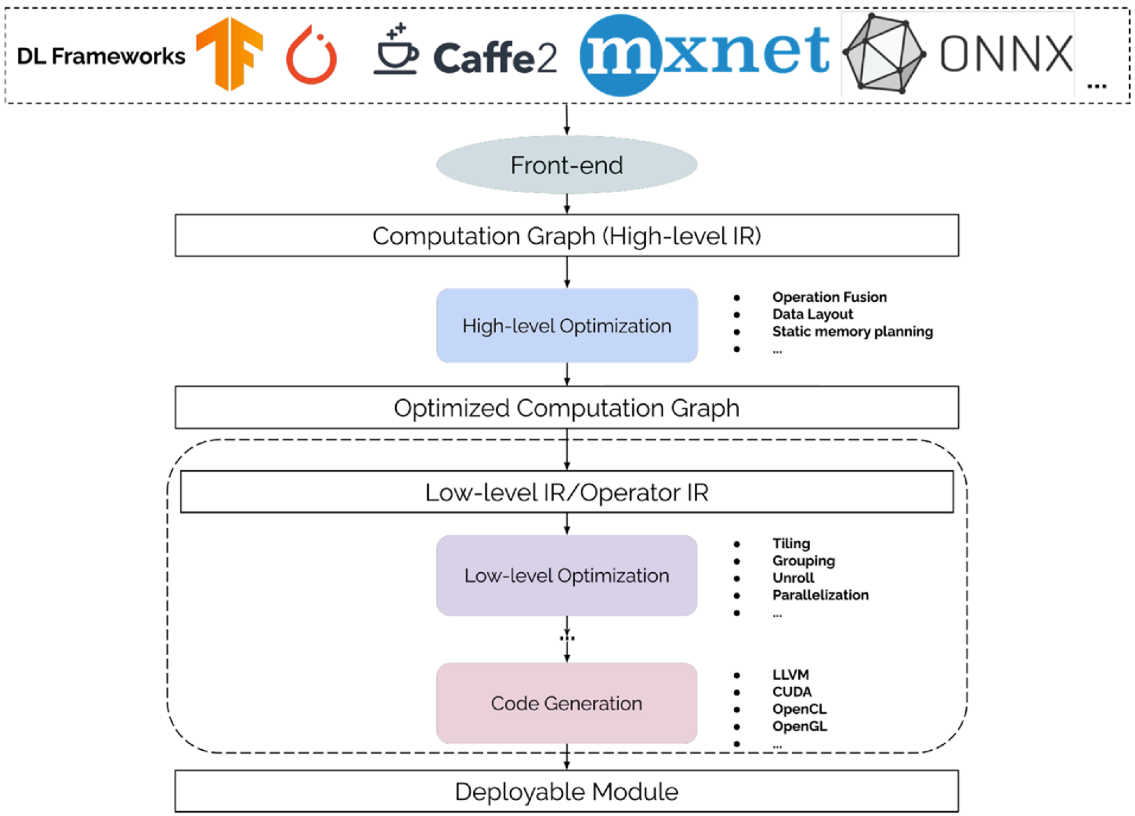
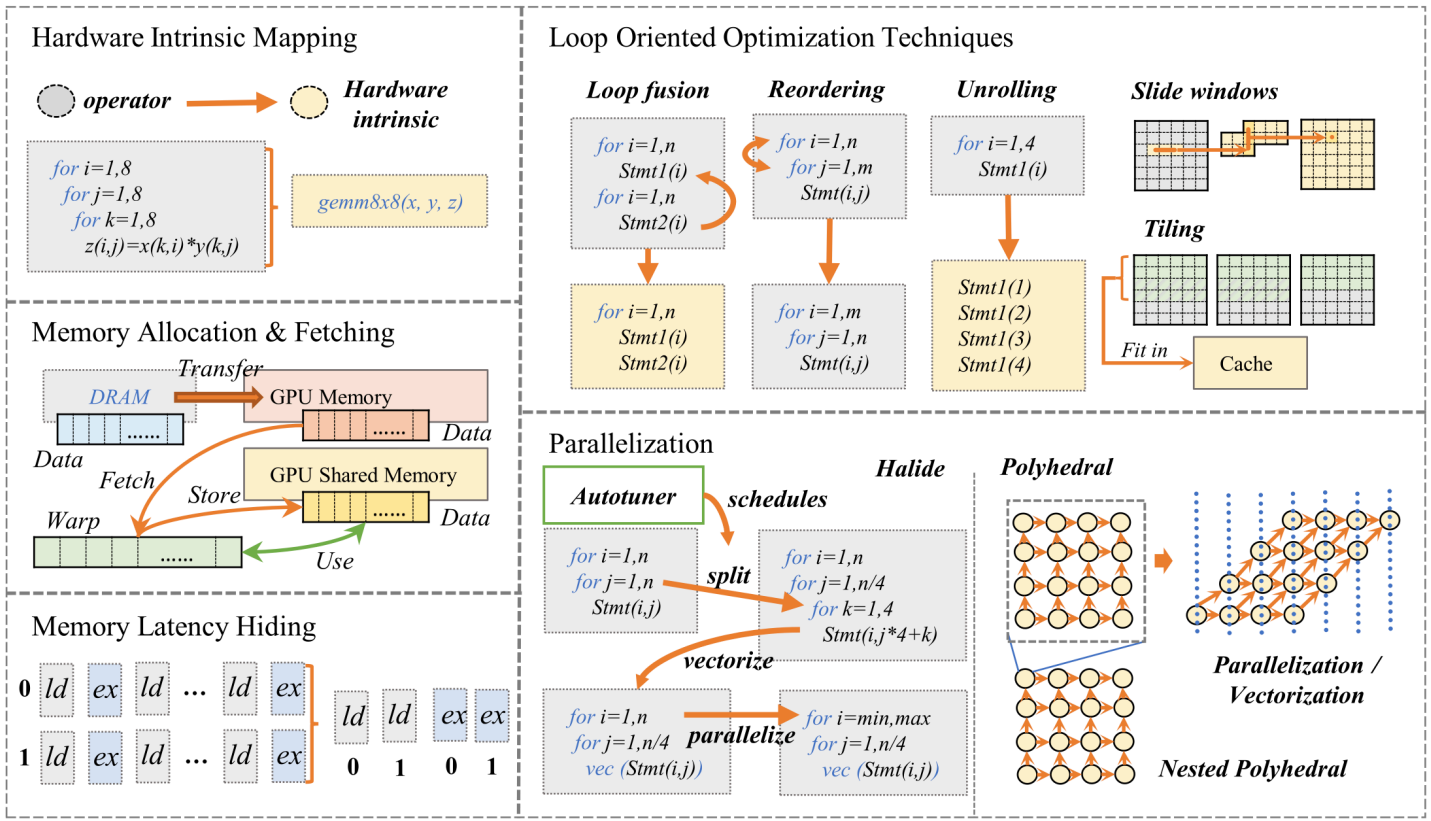
Machine Learning Framework are Just Optimizing Compilers



Machine Learning Framework (Cont.)



Li *et al.*, CSUR 2020, [3]

Peephole Optimizations (cf. [4], [5])

```asm
; x = x * 2
LOAD R1, 0      ; load from 0
MUL R1, 2       ; multiply R1 by 2
STORE R1, 0     ; store R1 back to 0
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```asm
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ADD R1, 0       ; add 0 to R1
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```asm
; x = x + 0
LOAD R1, 0
ADD R1, 0       ; add 0 to R1
STORE R1, 0
```

The optimized version removes the **unnecessary** addition operation.

⇓

```asm
LOAD R1, 0
STORE R1, 0
```

Loop Nest Optimizations — Loop Tiling (cf. [6], [7])

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

The vector `b` **may not** fit into a line of CPU cache, causing multiple cache misses during the inner loop.

It implies multiple **fetches** from the main memory, which is **slow**.

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```cpp
int TS = 16; // Tile Size
for (int jj=0; jj<m; jj+=TS) {
  for (int i=0; i<n; ++i) {
      for (int j=jj; j<MIN(jj + TS, m); ++j) {
          c[i][j] = a[i] * b[j];
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The inner loop works on a **tile** of `b` that fits into the cache.

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      }
  }
}
```

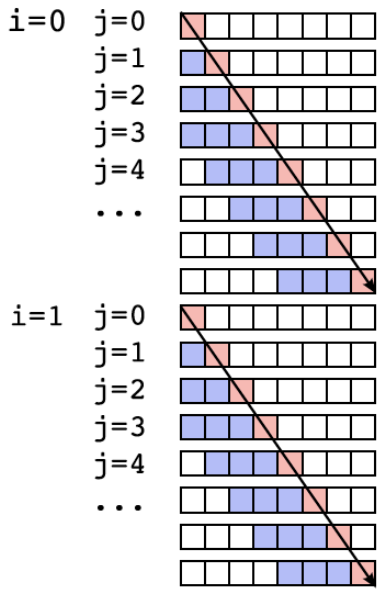
The inner loop works on a **tile** of `b` that fits into the cache.

But, the values for the array `a` will be read `m / TS` times!

Loop Tiling Visualization

Original:

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



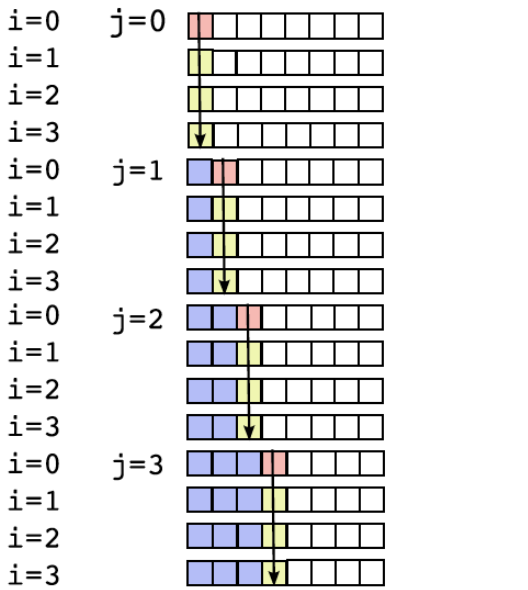
- cached, LRU eviction policy
- cache miss (read from memory, slow)
- cache hit (read from cache, fast)

Cache size: 4
TILE=4
(must be tuned to cache size)

Cache hit rate without tiling: 0%
Cache hit rate with tiling: 50%

Tiled:

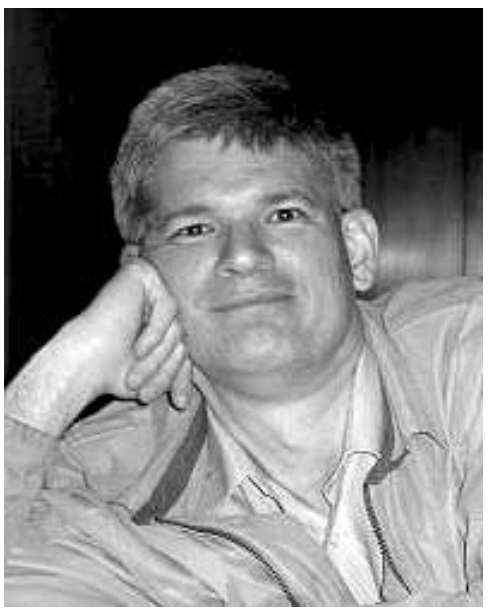
```
for (ii=0; ii<m; ii+=TILE)
  for (j=0; j<n; j++)
    for (i=ii; i<ii+TILE; i++)
      ...=...*b[j];
```



Tail Call/Recursion Optimization (cf. [8], [9], [10])

Guy L. Steele, Jr. in 1977 observed that **tail-recursive procedure calls** can be optimized to avoid growing the call stack [11]:

In general, procedure calls may be usefully thought of as GOTO statements which also pass parameters, and can be uniformly coded as [machine code] JUMP instructions.



Tail-recursive Factorial Function

C++

```cpp
int fact(int n) {
    if (n == 0) {
        return 1;
    }
    return n * fact(n - 1);
}
```

```
define i32 @fact(int)(i32 %n)  {
entry:
  %cmp3 = icmp eq i32 %n, 0
  br i1 %cmp3, label %return, label %if.else ; label %if.else.preheader
; if.else.preheader -> vector.ph -> vector.body -> middle.block -> if.else.preheader7
if.else:
  %n.tr5 = phi i32 [ %sub, %if.else ], [ %n.tr5.ph, %if.else.preheader7 ]
  %acc.tr4 = phi i32 [ %mul, %if.else ], [ %acc.tr4.ph, %if.else.preheader7 ]
  %sub = add nsw i32 %n.tr5, -1
  %mul = mul nsw i32 %n.tr5, %acc.tr4
  %cmp = icmp eq i32 %sub, 0
  br i1 %cmp, label %return, label %if.else
return:
  %acc.tr.lcssa = phi i32 [ 1, %entry ], [ %4, %middle.block ], [ %mul, %if.else ]
  ret i32 %acc.tr.lcssa
}
```

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