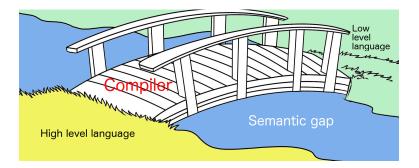
Compilers and computer architecture: Just-in-time compilation

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December 2019

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Recall the function of compilers



Welcome to the cutting edge

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Compilers are used to translate from programming languages humans can understand to machine code executable by computers. Compilers come in two forms:

- Conventional ahead-of-time compilers where translation is done once, long before program execution.
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- Just-in-time (JIT) compilers where translation of program fragments happens at the last possible moment and is interleaved with program execution.

We spend the whole term learning about the former. Today I want to give you a **brief** introduction to the latter.

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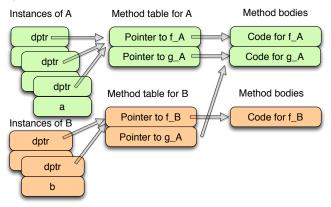
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JIT compilers are cutting (bleeding) edge technology and considerably more complex than normal compilers, which are already non-trivial. Hence the presentation today will be massively simplifying.

Let's look at two examples. Remember the compilation of objects and classes?



To deal with inheritance of methods, invoking a method is indirect via the method table. Each invocation has to follow two pointers. Without inheritance, no need for indirection.

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Of course an individual indirection takes < 1 nano-second on a modern CPU. So why worry? Answer: loops!

```
interface I {
  int f ( int n ); }
class A implements I {
  public int f ( int n ) { return n; } }
class B implements I {
  public int f ( int n ) { return 2*n; } }
class Main {
  public static void main ( String [] args ) {
  I \circ = new A ();
  for (int i = 0; i < 1000000; i++) {
    for (int j = 0; i < 1000000; j++) {
      o.f (i+j); } } }
```

Performance penalties add up.

But, I hear you say, it's obvious, even at compile time, that the object \circ is of class A. A good optimising compiler should be able to work this out, and replace the indirect invocation of f with a cheaper direct jump.

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Yes, in this simple example, a good optimising compiler can do this. But what about the following?

```
public static void main ( String [] args ) {
    I o = null;
    if ( args [ 0 ] == "hello" )
        new A ();
    else
        new B ();
    for ( int i = 0; i < 1000000; i++ ) {
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(Aside, can you see a hack to deal with this problem in an AOT compiler?)

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If JIT compilers are the answer ... what is the problem? Dynamically typed languages have a worse problem.

Simplifying a little, variables in dynamically typed languages store not just the usual value, e.g. 3, but also the type of the value, e.g. Int, and sometimes even more.

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Simplifying a little, variables in dynamically typed languages store not just the usual value, e.g. 3, but also the type of the value, e.g. Int, and sometimes even more. Whenever you carry an innocent operation like

x = x + y

under the hood something like the following happens.

```
let tx = typeof ( x )
let ty = typeof ( y )
if ( tx == Int && ty == Int )
    let vx = value ( x )
    let vy = value ( y )
    let res = integer_addition ( vx, vy )
    x_result_part = res
    x_type_part = Int
else
    ... // even more complicated.
```

Imagine this in a nested loop!

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This is painful. This is why dynamically typed languages are slow(er).

But ...

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So the compiler should move the type-checks outside the loops.

Recall that in dynamically typed languages

```
for ( int i = 0; i < 1000000; i++ ) {
  for ( int j = 0; i < 1000000; j++ ) {
    a [i, j] = a[i, j] + 1 }
}</pre>
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Is really

```
for ( int i = 0; i < 1000000; i++ ) {
  for ( int j = 0; i < 1000000; j++ ) {
    let ta = typeof ( a[i, j] ) // always same
    let t1 = typeof ( 1 ) // always same
    if ( ta == Int && t1 == Int ) {
        let va = value ( a[i, j] )
        let v1 = value ( 1 ) // simplifying
        let res = integer_addition ( va, v1 )
        a[ i, j ]_result_part = res
        a[ i, j ] _type_part = Int }
    else { ... } }</pre>
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So program from last slide can be

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let ta = typeof ( a )
let t1 = typeof ( 1 )
if ( ta == Array [...] of Int && t1 == Int ) {
for ( int i = 0; i < 1000000; i++ ) {
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Alas, at compile-time, the compiler does not have enough information to make this optimisation safely.

Let's summarise the situation.

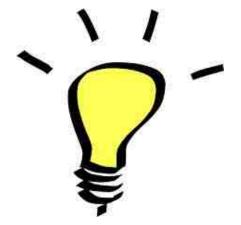
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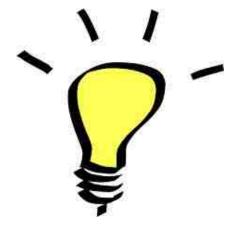
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Hmmm, what could we do ...



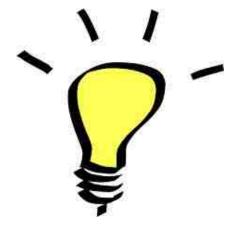


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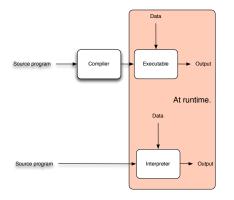


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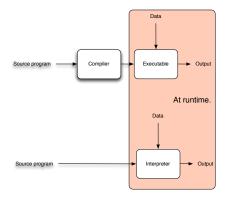
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Enter interpreters!

Recall from the beginning of the course, that interpreters are a second way to run programs.

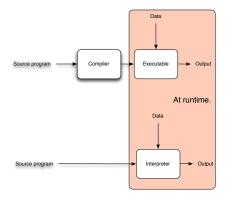


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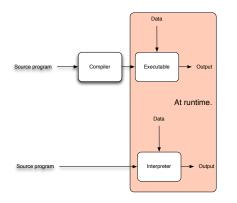
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- Compilers generate a program that has an effect on the world.
- Interpreters effect the world directly.

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- The advantage of compilers is that generated code is faster, because a lot of work has to be done only once (e.g. lexing, parsing, type-checking, optimisation). And the results of this work are shared in every execution. The interpreter has to redo this work every time.
- The advantage of interpreters is that they are much simpler than compilers.

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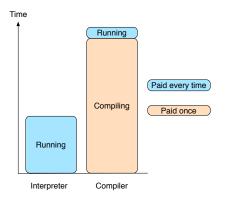
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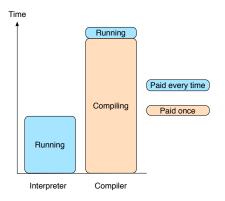
Hmmmm ...

Pareto principle and compiler/interpreter Δ to our rescue



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Combine this with the Pareto principle, and you have a potent weapon at hand.

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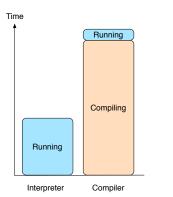
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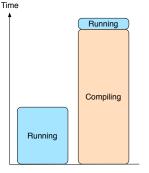
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Such code is referred to as hot.

Clearly compiling at run-time code that's executed infrequently will slow down execution. Trade-offs are different for hot code.



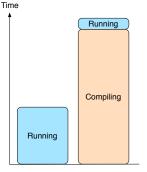
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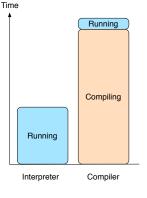


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Pareto's principle tells us that (typically) a program contains some hot code.

With the information available at run-time, we can aggressively optimise such hot code, and get a massive speed-up. The rest is interpreted. Sluggishness of interpretation doesn't matter, because it's only a fraction of program execution time.

There is just one problem ... how do we find hot code?

(reliably).



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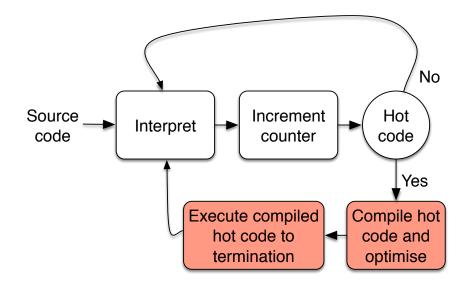
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When the compiled code terminates, we switch back to interpretation.

In a picture

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Aside

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Eventually, the hot code is detected and compiled with a great deal of optimisation. Then execution gets really fast.

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How actually to do the optimisations, taking information available at run-time into account.

Etc etc.

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So what's next in compiler technology? Let me introduce you to ...

Tracing JIT compilers are a form of JIT compilation where optimisation is especially aggressive.

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```
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  for ( y = 1 to 1000000 )
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Clearly the try-catch block is an innermost loop, so potentially hot code. But if the programmer does a good job, the exception handling will never be triggered. Yet we have all this exception handling code (tends to be large) in the hot loop. This causes all manner of problems, e.g. cache locality is destroyed.

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Why can't we use counters? Yes but ... counters only give us some relevant information ... for good optimisation we need more information. **Traces** give us this information. What are traces?

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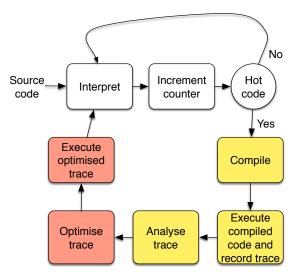
Once tracing has finished, e.g. the body of the hot loop has been executed once: then analyse and optimise trace.

Based on the analysis another compiler generates another (highly optimised) executable, which is then run to termination, then control goes back to interpreter.

Analysing and optimising the trace:

- Find out if variables change type in the loop, if not, move type-checking out of the loop. (For dynamically typed languages.)
- Find out if object change type in the loop, if not, use short-cut method invocations, no need to go via method table.
- Let the interpreter handle the rarely used parts of the hot loop (e.g. error handling).
- <u>ا...</u>
- Finally, enter the third phase, the 'normal' execution of the optimised trace.

A tracing JIT compiler in a picture



Difficulties

As with normal JIT compilers, we have to orchestrate the interplay of all these compiler phases, e.g.: Handover of control from interpreter to compiler, to tracing, to execution of optimised trace, and back. Garbage collection, exceptions, concurrency etc must all also work.

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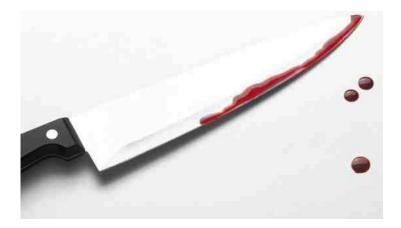
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Hard to say exactly who uses what (e.g. Apple Safari) since companies rarely say what they're using. They can use more than one. Trade secrets.

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Open source: **PyPy**, a **meta-tracing** framework for Python. **Meta-tracing**, what's that?





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The problem is that interpreter loops are the kinds of loops that JITers do not optimise well. Let's explain this in detail.

An interpreter is a big loop that gets the next command and acts on it, e.g.

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Now JIT compilers are really good at optimising loops, why do they fail with interpreter loops?

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Last requirement is violated in interpreter loops.

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This gives rise to something like the following bytecode

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loop:
    br r17 exit
    add r21 r33 r21
    subabs r33 1 r33
    jump loop
exit:
    ...
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Let's have bytecode and bytecode interpreter side-by-side:

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```
while true:
    op = mem [ pc ]
    pc = pc+1
    case op = br:
    r = mem [ pc ]
    pc = pc+1
    if mem [ r ] == 0:
        pc := mem [ pc ]
    case op = add:
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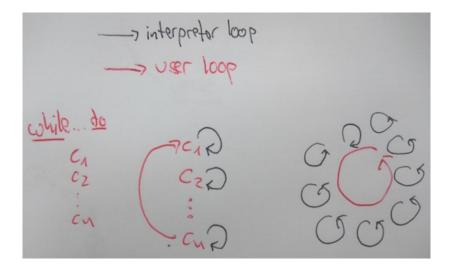
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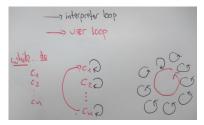
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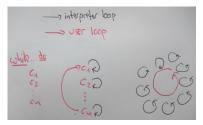
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Now every round of the interpreter takes a different branch. The tracing JIT can just optimise one branch through the loop. This is the worst case scenario: we pay the price of tracing, optimisation (since loop is executed a lot), only to throw away the optimisation and go back to interpretation.



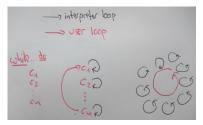


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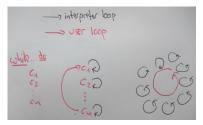
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This is too difficult to detect for profiling, since user programs can vary greatly.

The interpreter writer knows what the user loops are like:

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The idea of meta-tracing is to let the interpreter writer **annotate** the interpreter code with 'hooks' that tell the tracing JIT compiler where user loops start and end. The profiler can then identify the hot loops in (the interpretation of) user code.

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The real advantage of this is that it divides the problem of developing high-performance JIT compilers for a language into several parts, each of which separately is much more mangable:

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The PyPy meta-tracing framework runs Python substantially faster than e.g. the CPython framework.

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For bare-bones languages like C, there is little to optimise at run-time, and code generated by a conventional C compiler with heavy (hence slow) optimisation will almost always beat a modern JIT compiler.

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Much work left to be done.

Interested?

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Feel free to talk to me about this.



