

Incrementalization: From Clarity To Efficiency

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At the center of computer science

two major concerns of study:

what to compute

how to compute efficiently

problem solving:

from **clear** specifications for "what"

to **efficient** implementations for "how"

From clear specifications to efficient implementations

challenge:

develop a method that is both general and systematic

conflict between clarity and efficiency:

clear specifications usually correspond to straightforward implementations, not at all efficient.

efficient implementations are usually difficult to understand, not at all clear.

A general and systematic method

iterate: determine a minimum increment to take repeatedly, iteratively, to arrive at the desired program output

incrementalize: make expensive operations incremental in each iteration by using and maintaining useful additional values

implement: design appropriate data structures for efficiently storing and accessing the values maintained

applies to different programming paradigms	abstraction
loops: incrementalize	none
sets: incrementalize, implement	data
recursion: iterate , incrementalize	control
rules: iterate , incrementalize, implement	both
objects: incrementalize across components	module

iterate and incrementalize → integration by differentiation

Loops — a simple example

eliminating multiplications:

```
i:=1          -- in grid with a columns and b rows
while i <= b:
  :
  ...a*i...   -- access last element of each row
  :
  i:=i+1
```

strength reduction: an oldest optimization, for array access.

Difference Engine, ENIAC: tabulating polynomials.

need to use language semantics and cost model

exploit algebraic properties: $a*(i+1) = a*i+a$

store, update, initialize value of $a*i$: where? how?

Loops — incrementalize

incrementalize

maintain invariant: $c = a*i$, use and update

```
i:=1           → i:=1; c:=a;
while i <= b:
  :
  ...a*i...    → ...c...
  :
  i:=i+1       → i:=i+1; c:=c+a;
```

exploit algebraic properties

maintain additional information

iterate and implement: too little or too much to do

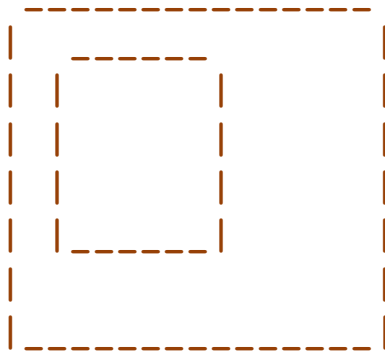
Loops — more examples

hardware design: non-restoring binary integer square root

```
n := input()
m := 2^(l-1)
for i := l-2 downto 0:
  p := n - m^2
  if p > 0:
    m := m + 2^i
  else if p < 0:
    m := m - 2^i
output(m)
```

goal: a few +- and shifts per bit

image processing: blurring



goal: a few operations per pixel

need higher-level abstraction

Sets — a simple example

graph reachability: edges, source vertices \rightarrow reachable vertices

```
r := s
```

```
while exists x in e[r]-r:
```

```
  r := r U {x}
```

need to

handle composite set expressions: $x[y]$, $x-y$

design representations of interrelated sets: e , s , r

Sets — incrementalize and implement

incrementalize: retrieve/add/delete element, test membership

two invariants for $e[r]$ - r : $t = e[r]$, $w = t - r$

chain rule: maintain t and then w .

derive rules for maintaining simple and complex invariants.

implement: s , domain of e , range of e , r , t , w

based representations: records for all elements of related sets;

a set retrieved from is a linked list of pointers to the records;

a set tested against is a field in the records.

iterate: directly from $\min r: s \text{ subset } r, r \cup e[r] = r$

Sets — more examples

query processing: join optimization

```
r := {[x,y]: x in s, y in t | f(x) = g(y)}
```

iterate:

```
r := {}  
for x in s:  
  r := r U {[x,y]: y in t | f(x)=g(y)}
```

incrementalize: maintain

```
ginverse = {[g(y),y]: y in t}
```

derived:

```
ginverse := {}  
for y in t:  
  ginverse = ginverse U {[g(y),y]}  
r := {}  
for x in s:  
  for y in ginverse{f(x)}  
    r := r U {[x,y]}
```

compare:

same asymptotic time: $O(\#s + \#t + \#r)$; fewer loops and ops;

less space: $O(\#t)$ or $O(\min(\#s, \#t))$, not $O(\#s + \#t)$; simpler and shorter; derived!

previous algorithm:

```
finverse := {}  
for x in s:  
  finverse := finverse U {[f(x),x]}  
ginverse := {}  
for y in t:  
  if g(y) in domain(finverse):  
    ginverse := ginverse U {[g(y),y]}  
r := {}  
for z in domain(ginverse):  
  for x in finverse{z}:  
    for y in ginverse{z}:  
      r := r U {[x,y]}
```

role-based access control (RBAC)

core RBAC: 16 expensive queries, 9 kinds, updated in many places.

125 lines python → hundreds of lines. CheckAccess: constant time.

Recursion — a simple example

longest common subsequence: sequences x and $y \rightarrow$ length

```
lcs(i,j)
= if i=0 or j=0: 0
  else if x[i]=y[j]: lcs(i-1,j-1)+1
  else: max(lcs(i,j-1),lcs(i-1,j))
```

need to

determine how to iterate: recursion to iteration

determine what and how to cache: dynamic programming

Recursion — iterate and incrementalize

```
lcs(i,j)
= if i=0 or j=0: 0
  else if x[i]=y[j]: lcs(i-1,j-1)+1
  else: max(lcs(i,j-1),lcs(i-1,j))
```

iterate: minimum increment from arguments of recursive calls
 $i, j \rightarrow i+1, j$

incrementalize: cache and use

```
lcs(i+1,j)    use r = lcs(i,j)           → lcs'(i,j,r)
= if i+1=0 lor j=0: 0
  else if x[i+1]=y[j]: lcs(i,j-1)+1    use lcs(i,j-1), cache
  else: max(lcs(i+1,j-1),lcs(i,j))    use lcs(i,j-1)
                                       → lcs'(i,j-1,lcs(i,j-1))
                                       recursively
```

implement: directly map to recursive or indexed data structures

Recursion — more examples

sequence processing: editing distance, paragraph formatting, matrix chain multiplications, ...

math puzzles: Hanoi tower, find solution in linear time

```
h(n,a,b,c)          -- move n disks from a to b using c
= if n<=0: skip
  else: h(n-1,a,c,b)::move(a,b)::h(n-1,c,b,a)
```

iterate: $n, a, b, c \rightarrow n+1, a, c, b$

cache: $\text{hExt}(n, a, b, c) = \langle \text{h}(n, a, b, c), \text{h}(n, b, c, a), \text{h}(n, c, a, b) \rangle$

```
hExt(n+1,a,c,b)    use rExt=hExt(n,a,b,c)  → hExt'(n,a,b,c,
= if n+1<=0: <skip,skip,skip>                rExt)
  else: 1st(rExt)::move(a,c)::2nd(rExt),
        3rd(rExt)::move(c,b)::1st(rExt),
        2nd(rExt)::move(b,a)::3rd(rExt)>
```

simpler than others: maintain 2 additional values, not 5

Rules — a simple example

transitive closure:

$\text{edge}(u,v) \rightarrow \text{path}(u,v)$

$\text{edge}(u,w), \text{path}(w,v) \rightarrow \text{path}(u,v)$

need to

find a way to proceed

determine what and how to maintain

design representations of different kinds of facts

additional question

can we give time and space complexity guarantees?

Rules — iterate, incrementalize, implement

iterate: add one fact at a time until fixed point is reached

incrementalize: maintain maps indexed by shared arguments

implement: design nested linked lists and arrays of records

time and space guarantees:

$\text{edge}(u,v) \rightarrow \text{path}(u,v)$

$\text{edge}(u,w), \text{path}(w,v) \rightarrow \text{path}(u,v)$

time: # of combinations of hypotheses — optimal

$O(\min(\#edge \times \#path^{2/1}, \#path \times \#edge^{1/2}))$

edges vertices output indegree

space: $O(\#edge)$, for storing inverse map of edge

Rules — more examples

program analysis: dependence analysis, pointer analysis, information flow analysis, ...

trust management: SPKI/SDSI authorization

```
auth(k1, [k2], TRUE, a1, v1), auth(k2, s2, d2, a2, v2)
  -> auth(k1, s2, d2, PInt(a1, a2), VInt(v1, v2))
```

```
auth(k1, [k2 [n2 ns3]], d, a, v1), name(k2, n2, [k3], v2)
  -> auth(k1, [k3 ns3], d, a, VInt(v1, v2))
```

```
name(k1, n1, [k2 [n2 ns3]], v1), name(k2, n2, [k3], v2)
  -> name(k1, n1, [k3 ns3], VInt(v1, v2))
```

find authorized keys: $O(in * kp * kn)$, better than $O(in * k * k)$.

Objects — a simple example

the “what” of a software component:

queries: compute information using data w/o changing data.

updates: change data.

example:

class `LinkedList` in Java has many methods:

`size()`, `add` or `remove`, several other queries.

Objects — incrementalize

how to implement the queries and updates: varies significantly

straightforward:

queries compute requested information.

updates change base data.

example: `size()` contains a loop that computes the size.

observe:

queries are often repeated, many are easily expensive;

updates can be frequent, they are usually small.

sophisticated — incrementalized:

store derived information; queries return stored information.

updates also update stored information.

example: maintain size in a field, and update it in 18 places.

Objects — more examples

examples: wireless protocols, electronic health records, virtual reality, games, ...

```
findStrongSignals(): return {s in signals | s.getStrength() > threshold}
```

```
class Protocol
  signals: set(Signal)
  threshold: float
+ strongSignals: set(Signal)
  ...
  addSignal(signal): signals.add(signal)
+   signal.takeProtocol(this)
+   if signal.getStrength() > threshold
+     strongSignals.add(signal)
* findStrongSignals(): return strongSignals
+ updateSignal(signal):
+   if signals.contains(signal)
+     if strongSignals.contains(signal)
+       if not signal.getStrength()>threshold
+         strongSingals.remove(signal)
+     else
+       if signal.getStrength()>threshold
+         strongSingals.add(signal)
  ...
```

```
class Signal
  strength: float
+ protocols: set(Protocol)
  ...
+ takeProtocol(protocol):
+   protocols.add(protocol)
  setStrength(v):
    strength = v
+   for protocol in protocols
+     protocol.updateSignal(this)
  getStrength(): return strength
  ...
  ...
```

original lines
* changed lines
+ added lines

```
findStrongSignal:  $O(\#signals) \rightarrow O(1)$ . setStrength:  $O(1) \rightarrow O(\#protocols)$ .
```

Iterate, Incrementalize, Implement

iterate at a minimum increment step; incrementalize expensive computations; implement on efficient data structures.

loops

iter, **inc**, impl

maintaining invariants, algebraic properties, additional values

sets

iter, **inc**, **impl**

chain rule, deriving maintenance rules; based representations

recursion

iter, **inc**, impl

recursion to iteration; dynamic programming

rules

iter, **inc**, **impl**

all, giving time and space complexity guarantees

objects

all, across components

connect theory w/ practice. like differentiation & integration.

References

loops [Allen69..., Liu97, LS98a/LSLR05...]

sets [Earley76..., PK82, Willard96, Willard02, LWGRCZZ06...]

recursion [BD77..., Smith90, LS99/03, LS00, LS02/09...]

rules [Forgy82, Vardi82..., McAllester99, LS03/09...]

objects [..., LSGRL05, RL08...]

more: *Systematic Program Design: From Clarity to Efficiency*

Ongoing projects

objects and nested sets: generating incremental implementations of queries with complexity guarantees

datalog and extensions: generating efficient implementations for demand-driven queries [PPDP 10, SIGMOD 11], functors, and arbitrary negation

logic quantifications: generating optimized implementations by reducing nesting levels and using aggregates [OOPSLA 12]

distributed algorithms: from very high-level specifications to efficient implementations [OOPSLA 12, SSS 12]