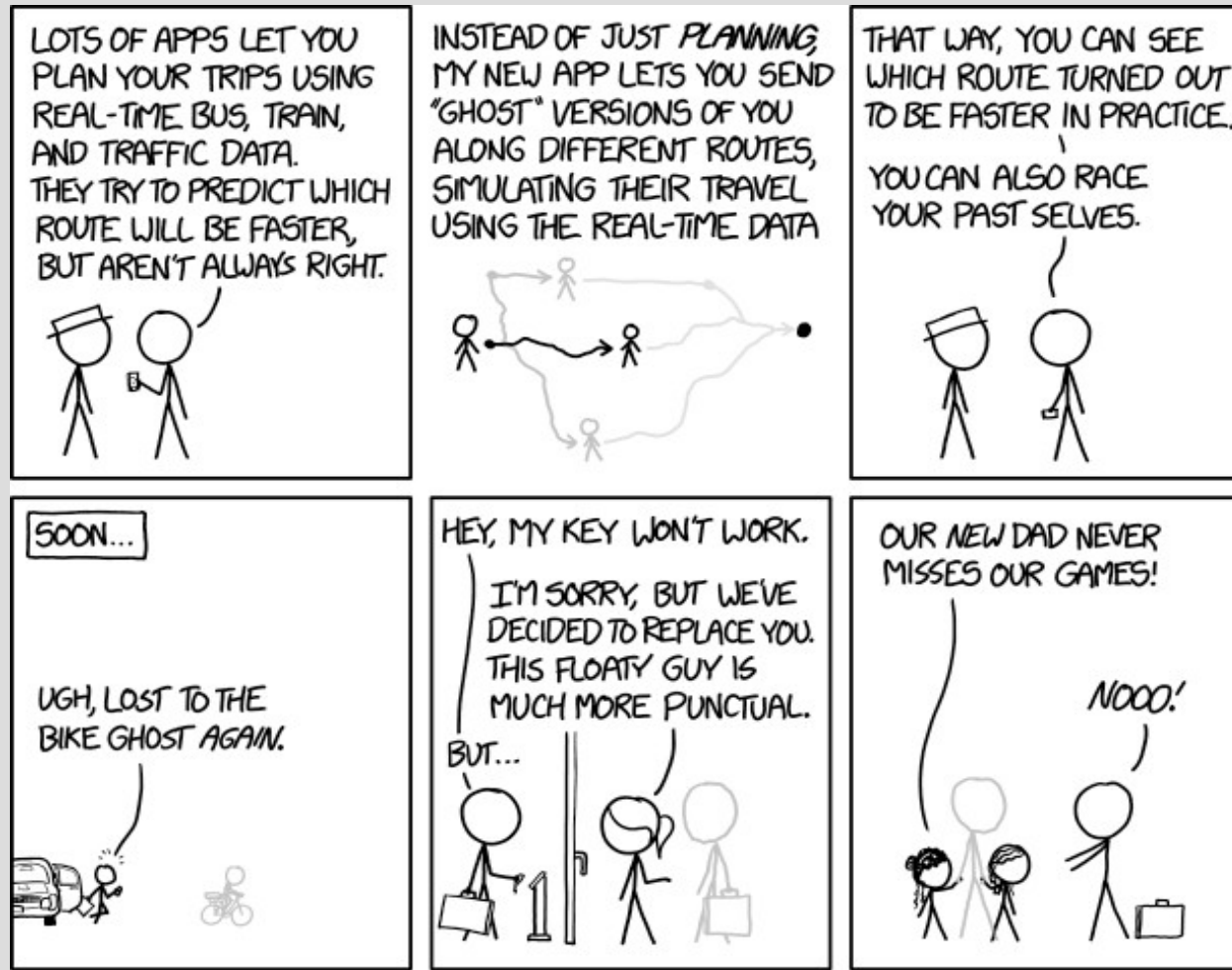


# All-Pairs Shortest Paths



# TL;DR dynamic programming

What are two ways you can compute the Fibonacci numbers?

$$F_n = F_{n-1} + F_{n-2}$$

with  $F_0 = 0, F_1 = 1$

Which way is better?

# TL;DR dynamic programming

One way, simply use the definition

Recursive:

$F(n)$ :

if( $n==1$  or  $n==0$ )

return  $n$

else return  $F(n-1)+F(n-2)$

# TL;DR dynamic programming

Another way, compute  $F(2)$ , then  $F(3)$   
... until you get to  $F(n)$

Bottom up:

$$A[0] = 0$$

$$A[1] = 1$$

for  $i = 2$  to  $n$

$$A[i] = A[i-1] + A[i-2]$$

# TL;DR dynamic programming

This second way is *much* faster

It turns out you can take pretty much any recursion and solve it this way (called “dynamic programming”)

It can use a bit more memory, but much faster

# TL;DR dynamic programming

How many multiplication operations does it take to compute:

$x^4$ ?

$x^{10}$ ?

# TL;DR dynamic programming

How many multiplication operations does it take to compute:

$x^4$ ? Answer: 2

$x^{10}$ ? Answer: 4

# TL;DR dynamic programming

Can compute  $x^4$  with 2 operations:

$x^2 = x * x$  (store this value)

$x^4 = x^2 * x^2$

Save CPU by using more memory!

Can compute  $x^n$  using  $O(\lg n)$  ops

Also true if  $x$  is a matrix



# Shortest paths using matrices

Any sub-path ( $p_{x,y}$ ) of a shortest path ( $p_{u,v}$ ) is also a shortest path



Thus we can recursively define a shortest path  $p_{0,k} = \langle v_0, \dots, v_k \rangle$ , as:

$$w(p_{0,k}) = \min_{k-1} (w(p_{0,k-1}) + w(\text{"k-1", k}))$$

# Shortest paths using matrices

Thus a shortest path (using less than  $m$  edges) can be defined as:

$$L^m = l^m_{i,j} = \min_k (l^{m-1}_{i,k} + l^1_{k,j}),$$

where  $L^1$  is the edge weights matrix

Can use dynamic programming to find an efficient solution

# Shortest paths using matrices

$L^m$  is not the  $m^{\text{th}}$  power of  $L$ , but the operations are very similar:

$$L^m = l^m_{i,j} = \min_k (l^{m-1}_{i,k} + l^1_{k,j}) \text{ // ours}$$

$$L^m = l^m_{i,j} = \sum_k (l^{m-1}_{i,k} * l^1_{k,j}) \text{ // real times}$$

Thus we can use our multiplication saving technique here too!

(see: MatrixAPSPmult.java)

# Shortest paths using matrices

All-pairs-shortest-paths( $W$ )

$L(1) = W$ ,  $n = W.rows$ ,  $m = 1$

while  $m < n$

$L(2m) = ESP(L(m), L(m))$

$m = 2m$

return  $L(m)$

(ESP is L min op on previous slide)

# Shortest paths using matrices

Runtime:

$$|V|^3 \lg |V|$$

Correctness:

By definition (brute force with some computation savers)

# Floyd-Warshall

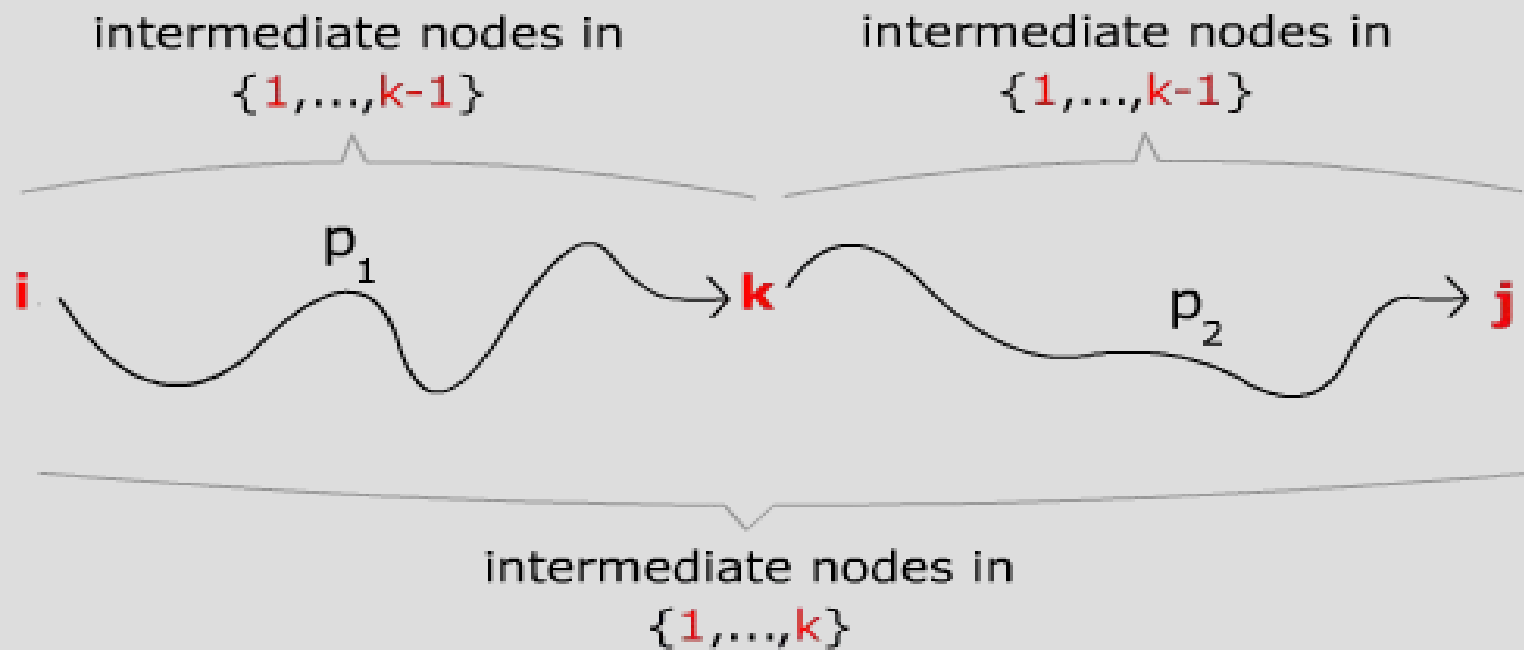
The Floyd-Warshall is similar but uses another shortest path property

Suppose we have a graph  $G$ , if we add a single vertex  $k$  to get  $G'$

We now need to recompute all shortest paths

# Floyd-Warshall

Either the path goes through  $k$ ,  
or remains unchanged



$$d_{i,j}^k = \min (d_{i,j}^{k-1}, d_{i,k}^{k-1} + d_{k,j}^{k-1})$$

# Floyd-Warshall

Floyd-Warshall(W) // dynamic prog

$d^0_{i,j} = W_{i,j}$ ,  $n = W.rows$

for  $k = 1$  to  $n$

  for  $i = 1$  to  $n$

    for  $j = 1$  to  $n$

$d^k_{i,j} = \min (d^{k-1}_{i,j}, d^{k-1}_{i,k} + d^{k-1}_{k,j})$



# Floyd-Warshall

Runtime:

$O(|V|^3)$

Correctness:

Again, by definition of shortest path