

CSE100: Design and Analysis of Algorithms

Lecture 19 – Weighted Graphs

Apr 5th 2022

Dijkstra and Bellman-Ford

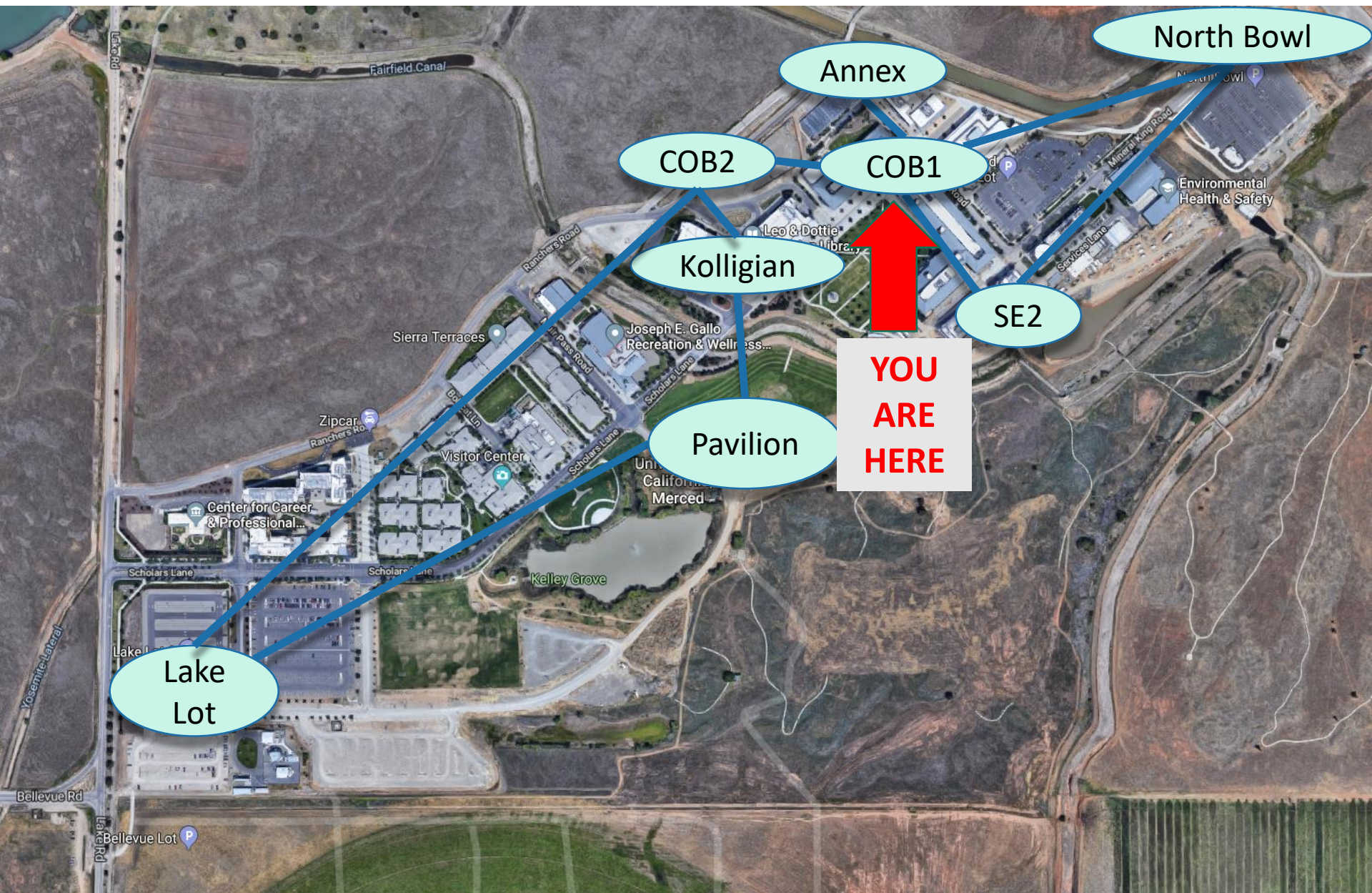


Today

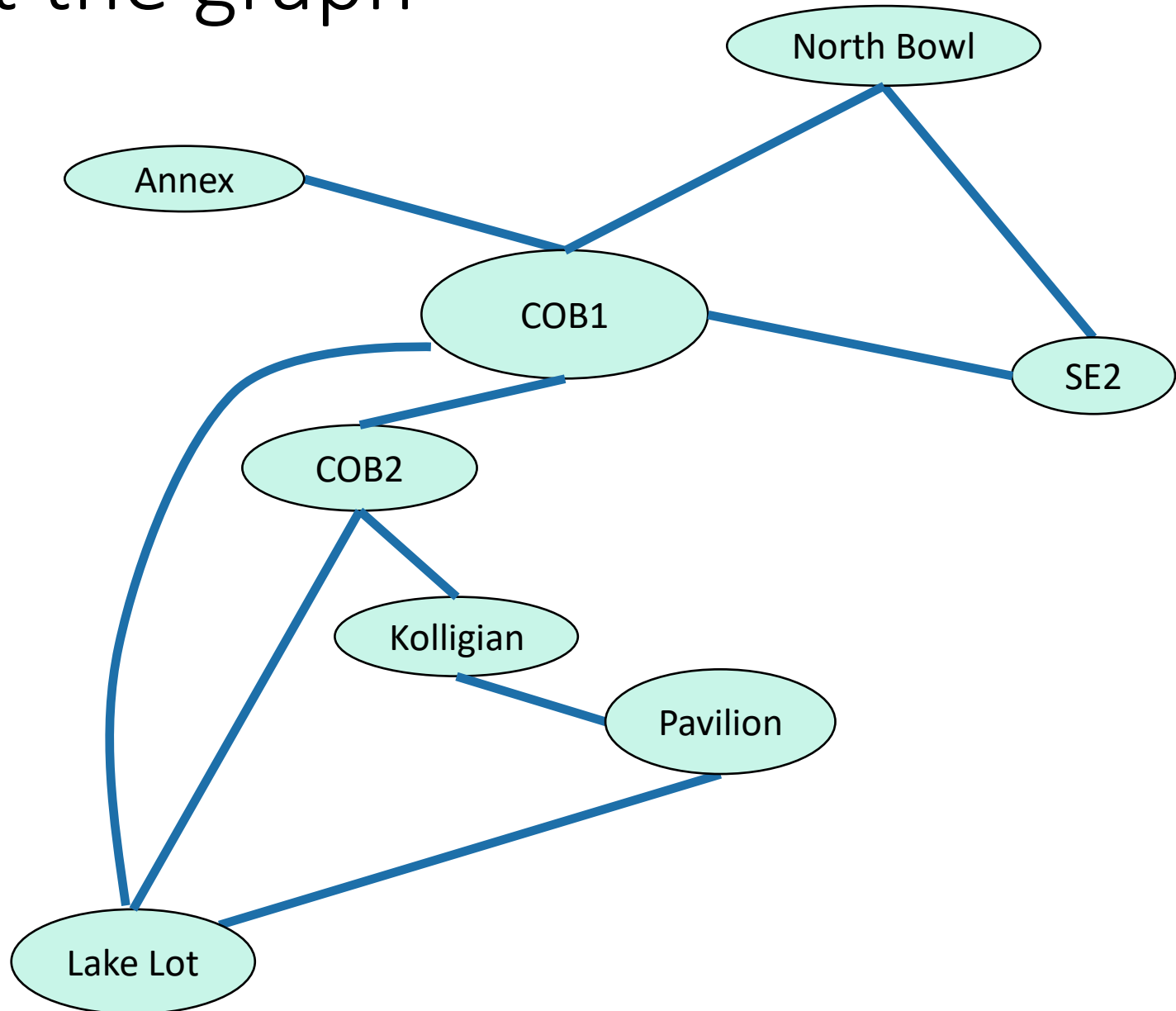


- What if the graphs are weighted?
- Part A: Dijkstra!
 - This will take most of today's class
- Part B: Bellman-Ford!
 - Real quick at the end!
 - We'll come back to Bellman-Ford in more detail, so today is just a taste.

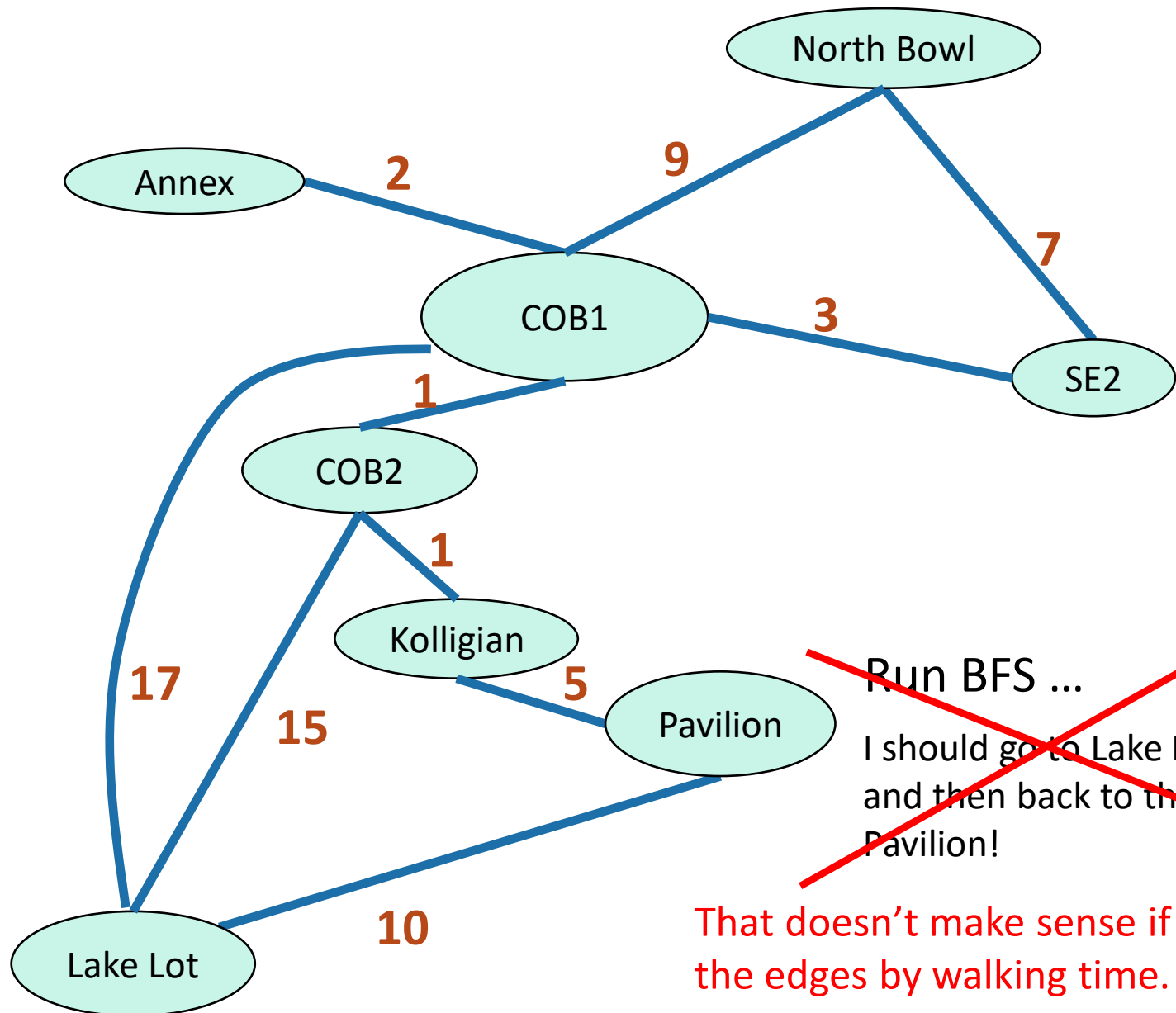




Just the graph



Shortest path from COB1 to Pavilion?



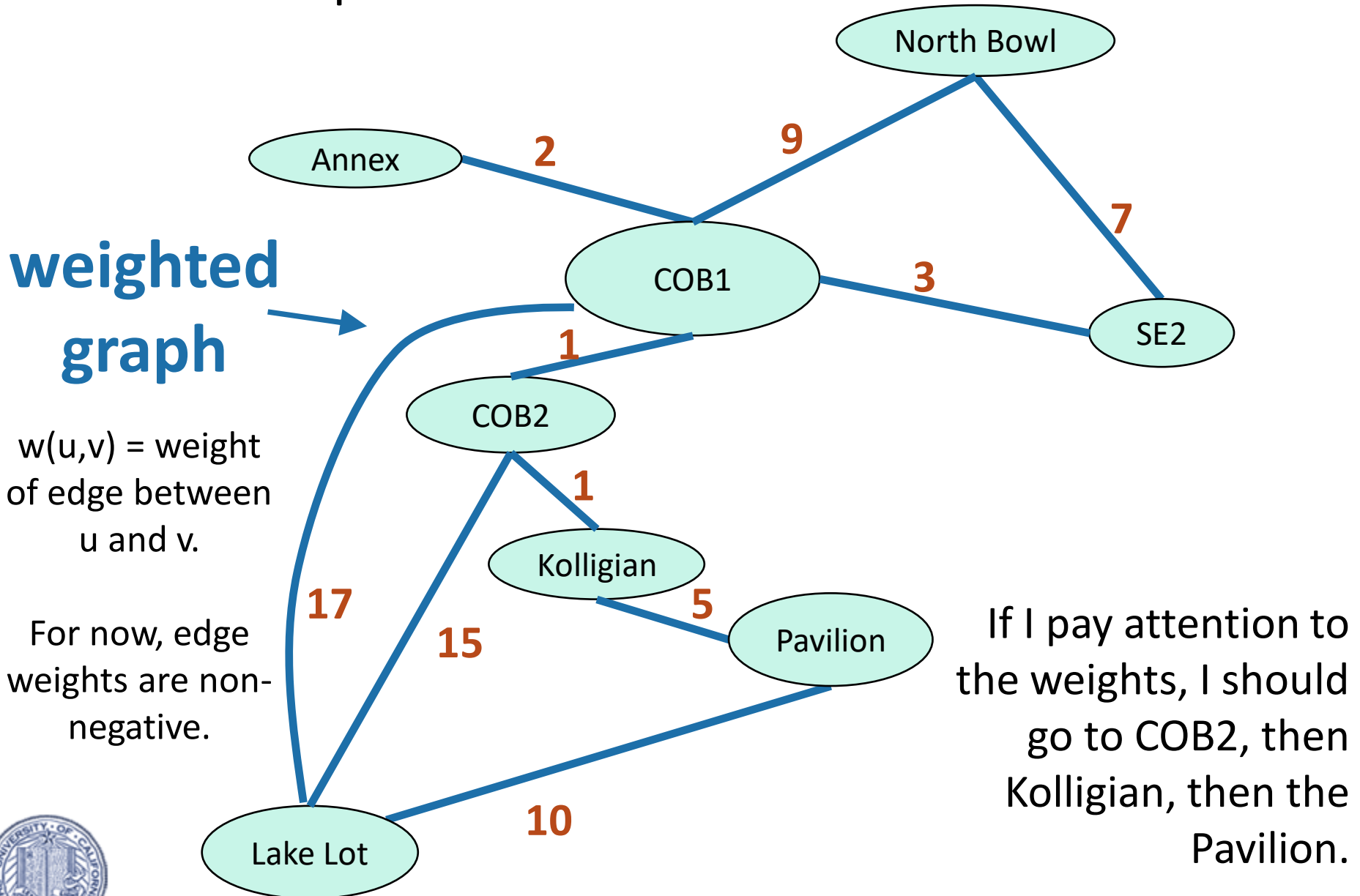
~~Run BFS ...~~

~~I should go to Lake Lot
and then back to the
Pavilion!~~

That doesn't make sense if I label
the edges by walking time.

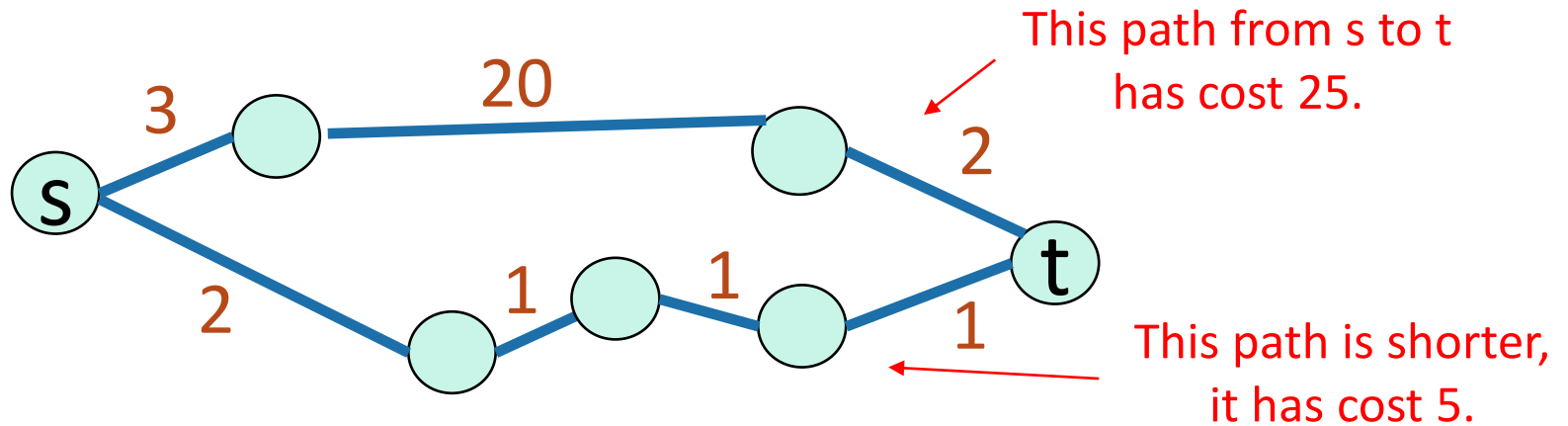


Shortest path from COB1 to the Pavilion?

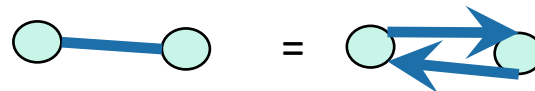


Shortest path problem

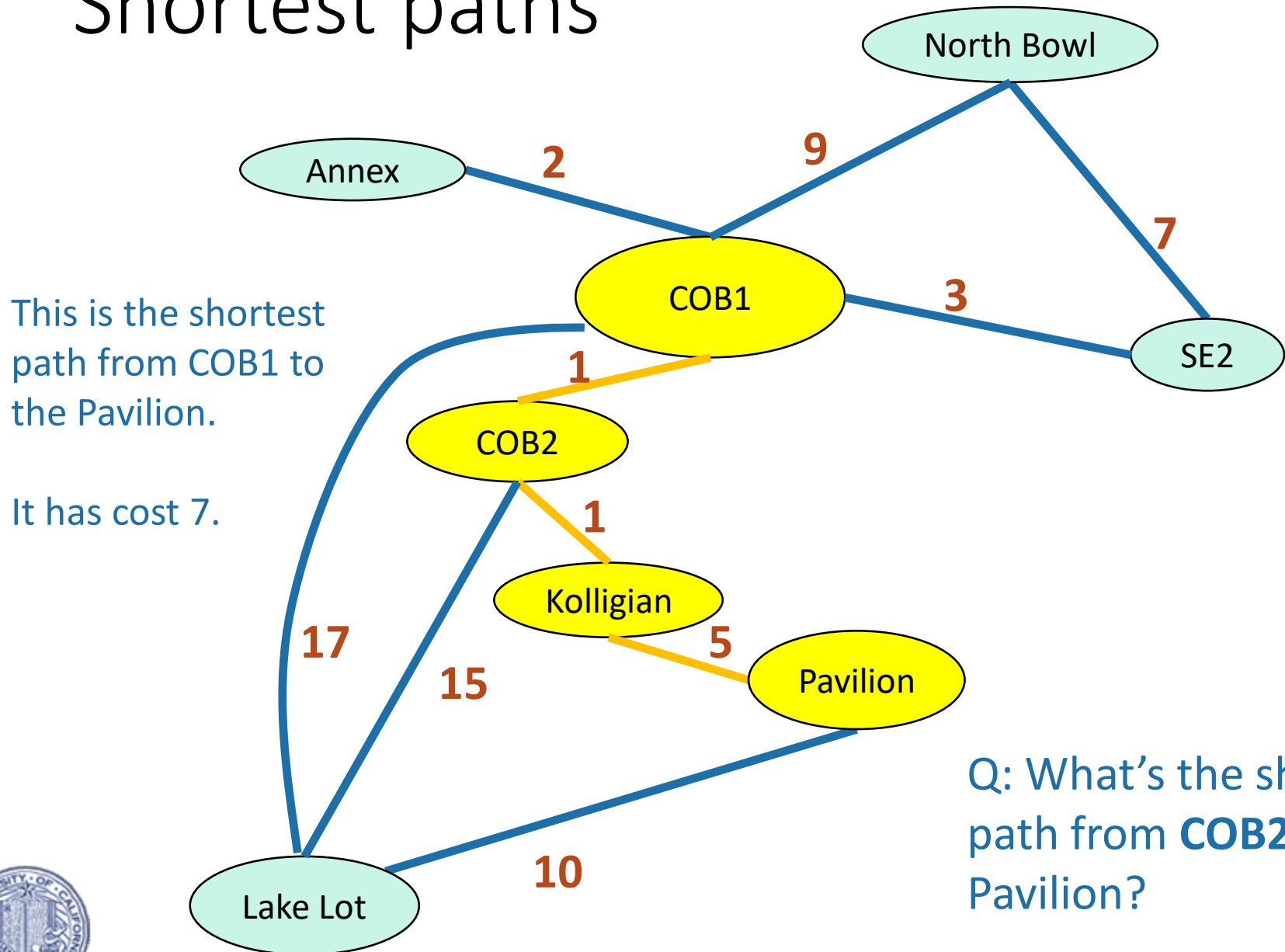
- What is the **shortest path** between u and v in a weighted graph?
 - the **cost** of a path is the sum of the weights along that path
 - The **shortest path** is the one with the minimum cost.



- The **distance** $d(u,v)$ between two vertices u and v is the cost of the shortest path between u and v .
- For this lecture **all graphs are directed**, but to save on notation I'm just going to draw undirected edges.



Shortest paths



Warm-up

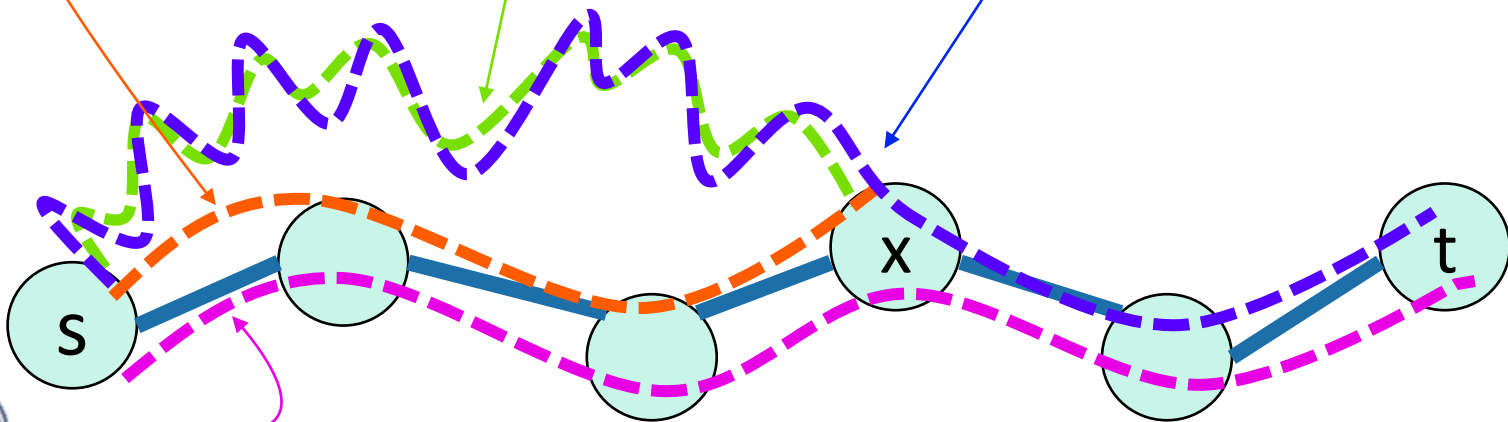
- A sub-path of a shortest path is also a shortest path.

- Say **this** is a shortest path from s to t .

- Claim: **this** is a shortest path from s to x .

- Suppose not, **this** one is a shorter path from s to x .
- But then that gives an **even shorter path** from s to t !

CONTRADICTION!!



Single-source shortest-path problem

- I want to know the shortest path from one vertex (COB1) to all other vertices.

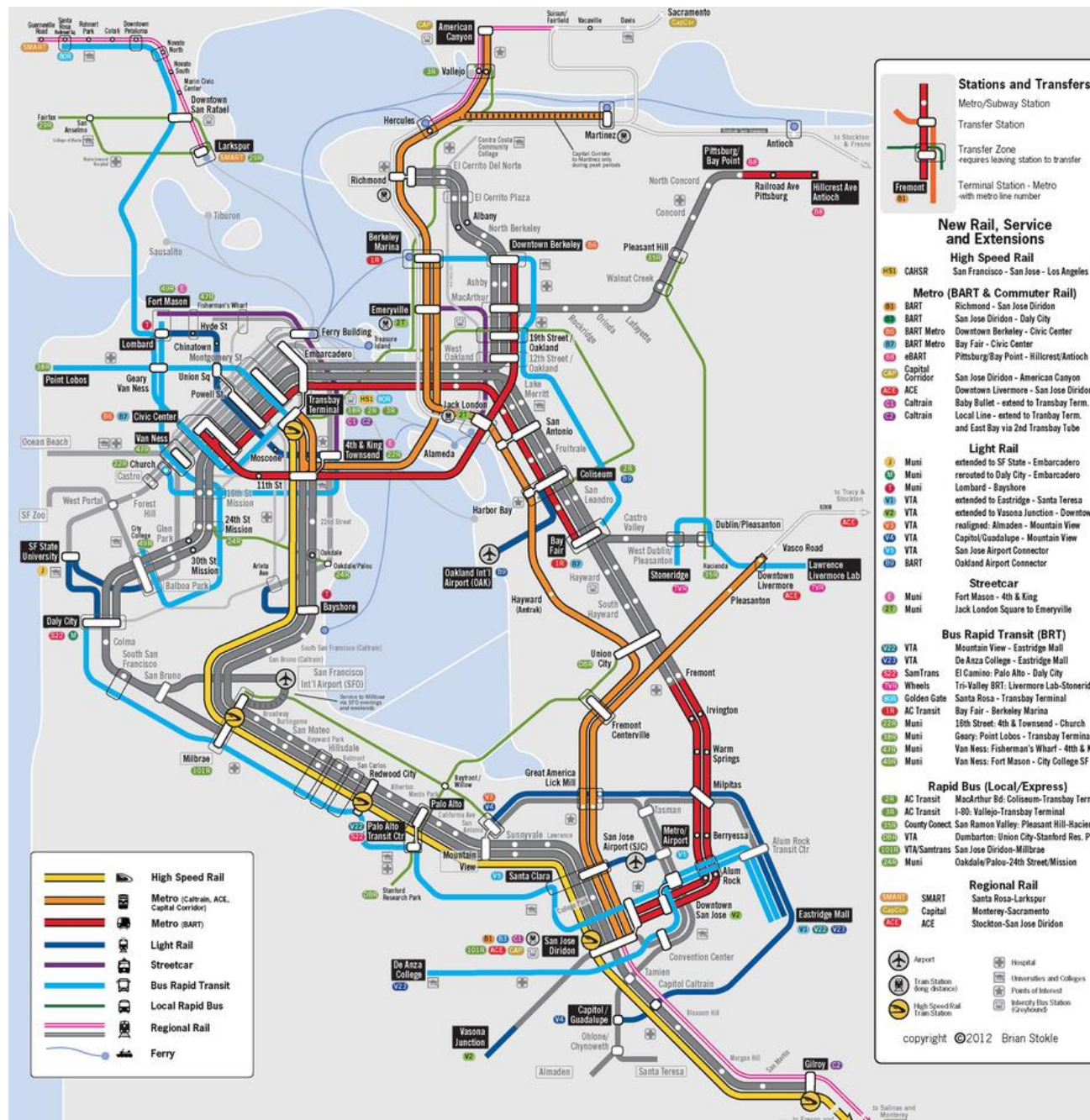
Destination	Cost	To get there
COB2	1	COB2
Kolligian	2	COB2-Kolligian
Annex	2	Annex
North Bowl	9	North Bowl
Pavilion	7	COB2-Kolligian-Pavilion
SE2	3	SE2
Lake Lot	16	COB2-Lake Lot

(Not necessarily stored as a table – how this information is represented will depend on the application)



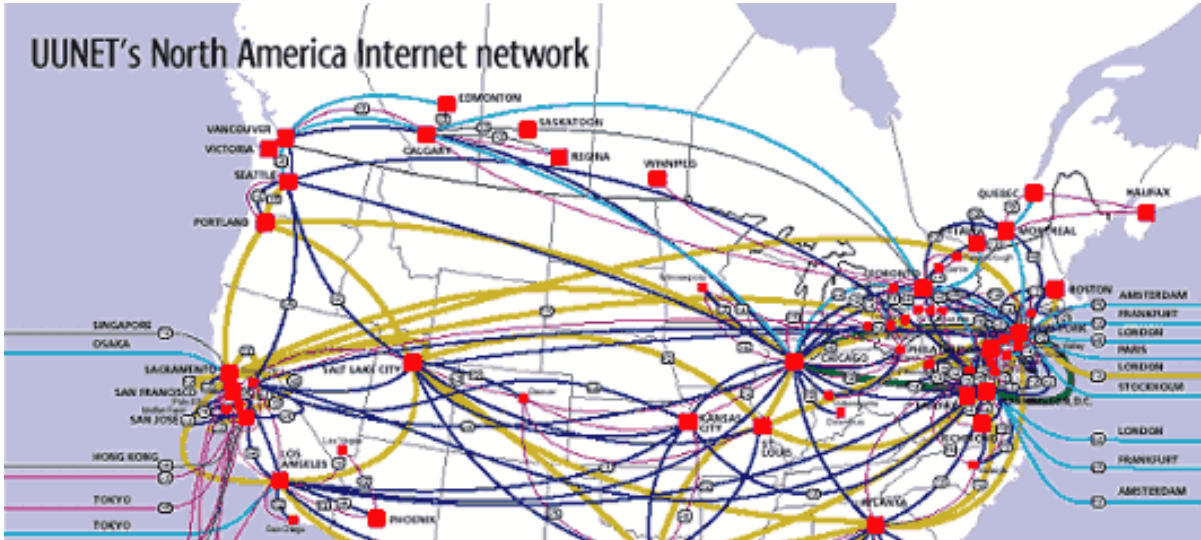
Example

- “what is the shortest path from Palo Alto to [anywhere else]” using BART, Caltrain, lightrail, MUNI, bus, Amtrak, bike, walking, uber/lyft.
- Edge weights have something to do with time, money, hassle.



Example

- **Network routing**
- I send information over the internet, from my computer to to all over the world.
- Each path has a cost which depends on link length, traffic, other costs, etc..
- How should we send packets?



```
7:33pm root@mars:~#[5]traceroute -A www.ethz.ch
traceroute to www.ethz.ch (129.132.19.216), 30 hops max, 60 byte packet
 1  169.236.151.2 (169.236.151.2) [AS22323]  0.339 ms  0.330 ms  0.330 ms
 2  10.7.1.177 (10.7.1.177) [*]  0.346 ms  0.307 ms  0.303 ms
 3  10.7.2.2 (10.7.2.2) [*]  0.861 ms  0.883 ms  0.859 ms
 4  10.7.2.18 (10.7.2.18) [*]  1.143 ms  1.067 ms  1.132 ms
 5  10.7.1.226 (10.7.1.226) [*]  11.272 ms  11.308 ms  11.338 ms
 6  hpr-tri-hpr3--ucm-10ge.cenic.net (137.164.27.113) [AS2152]  6.344 ms  6.344 ms  6.344 ms
 7  hpr-riv-hpr3--tri-hpr3-100g.cenic.net (137.164.25.93) [AS2152]  16.999 ms  17.022 ms  17.045 ms
 8  137.164.25.86 (137.164.25.86) [AS2152]  16.999 ms  17.022 ms  17.045 ms
 9  hpr-lax-hpr3--sdg-hpr3-100ge.cenic.net (137.164.25.90) [AS2152]  17.045 ms  17.045 ms  17.045 ms
10  hpr-i2--lax-hpr3-r-and-e.cenic.net (137.164.26.201) [AS2152]  17.045 ms  17.045 ms  17.045 ms
11  ae-5.4079.rts.wash.net.internet2.edu (162.252.70.158) [AS11537]  152.712 ms  159.212 ms  166.412 ms
12  internet2-gw.mx1.lon.uk.geant.net (62.40.124.44) [AS21320/AS20965]  152.712 ms  159.212 ms  166.412 ms
13  ae6.mx1.lon2.uk.geant.net (62.40.98.37) [AS21320/AS20965]  152.712 ms  159.212 ms  166.412 ms
14  ae5.mx1.par.fr.geant.net (62.40.98.179) [AS21320/AS20965]  152.712 ms  159.212 ms  166.412 ms
15  ae5.mx1.gen.ch.geant.net (62.40.98.182) [AS21320/AS20965]  152.712 ms  159.212 ms  166.412 ms
16  swice1-100ge-0-3-0-1.switch.ch (62.40.124.22) [AS21320/AS20965]  166.912 ms  169.068 ms  169.329 ms
17  swice4-100GE-0-0-0-0.switch.ch (130.59.36.6) [AS559]  166.912 ms  169.068 ms  169.329 ms
18  swibe3-100GE-0-1-0-1.switch.ch (130.59.37.145) [AS559]  166.912 ms  169.068 ms  169.329 ms
19  swibf1-100GE-0-0-0-1.switch.ch (130.59.39.78) [AS559]  166.912 ms  169.068 ms  169.329 ms
20  swiez3-100GE-0-1-0-0.switch.ch (130.59.37.6) [AS559]  166.912 ms  169.068 ms  169.329 ms
21  rou-gw-lee-tengig-to-switch.ethz.ch (192.33.92.1) [AS559]  170.666 ms  170.305 ms  170.305 ms
22  rou-fw-rz-rz-gw.ethz.ch (192.33.92.169) [AS559]  170.666 ms  170.305 ms  170.305 ms
23  www.ethz.ch (129.132.19.216) [AS559]  170.325 ms  170.305 ms  170.305 ms
```



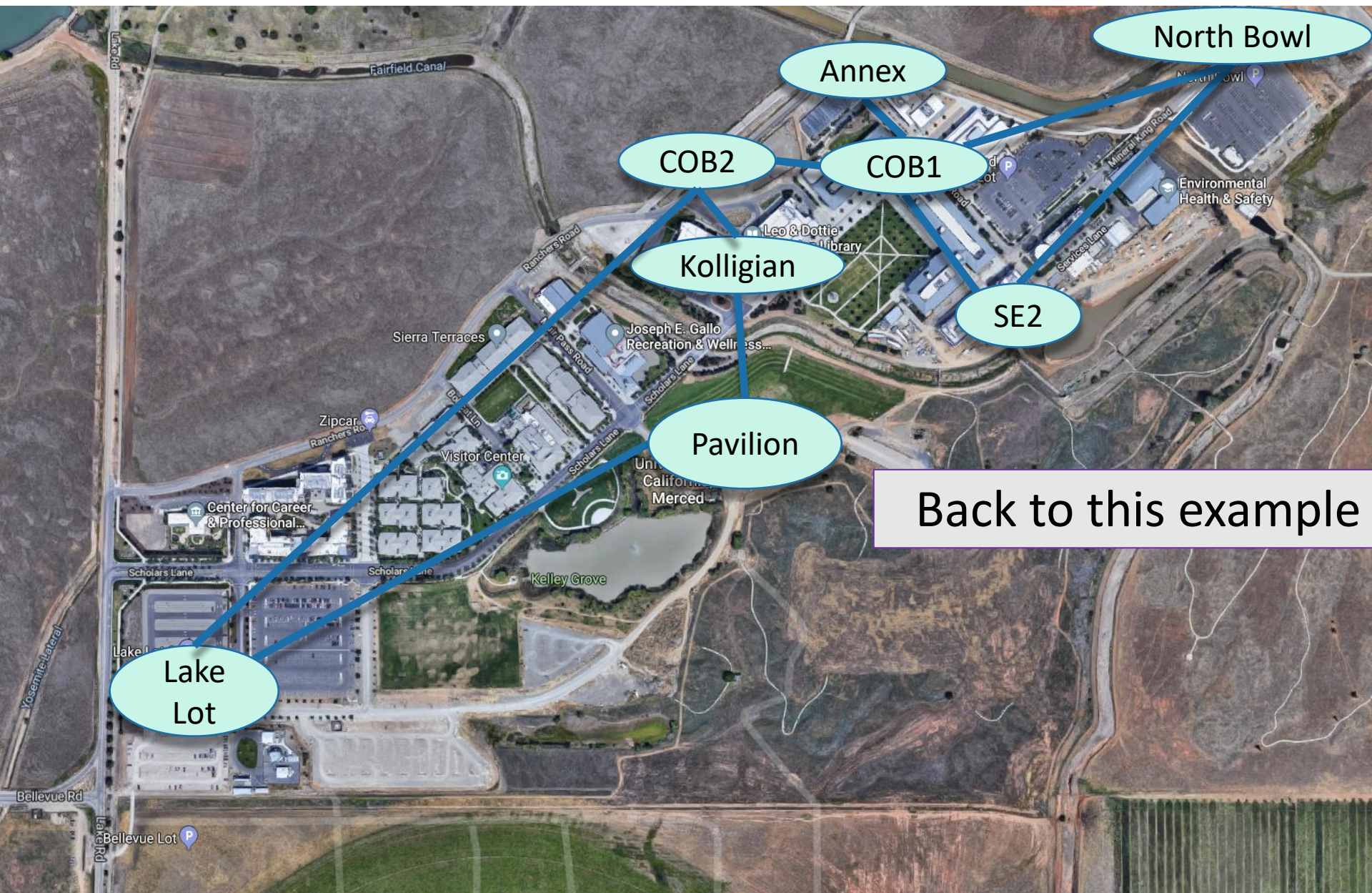
Aside: These are difficult problems

- Costs may change
 - If it's raining the cost of biking is higher
 - If a link is congested, the cost of routing a packet along it is higher
- The network might not be known
 - My computer doesn't store a map of the internet
- We want to do these tasks really quickly
 - I have time to bike to Savemart, but not to think about whether I should bike to Savemart...
 - More seriously, **the Internet.**

← This is a joke.

But let's ignore them for now.



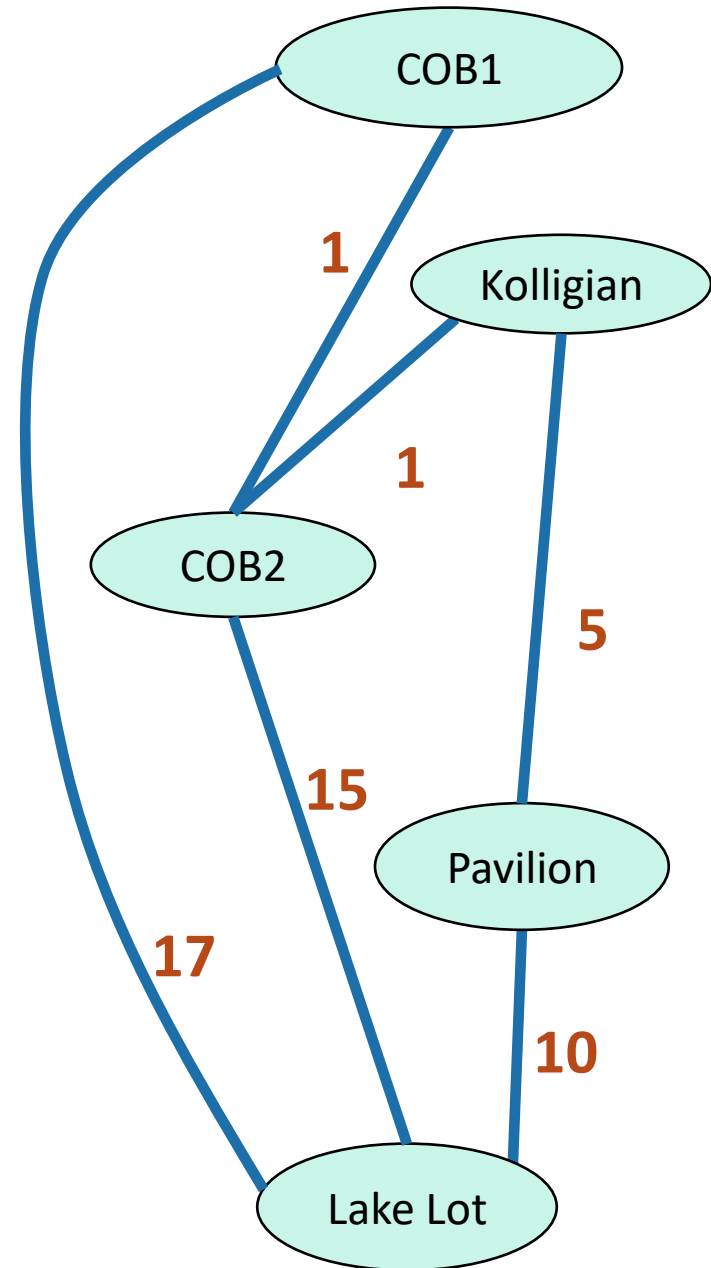


Back to this example



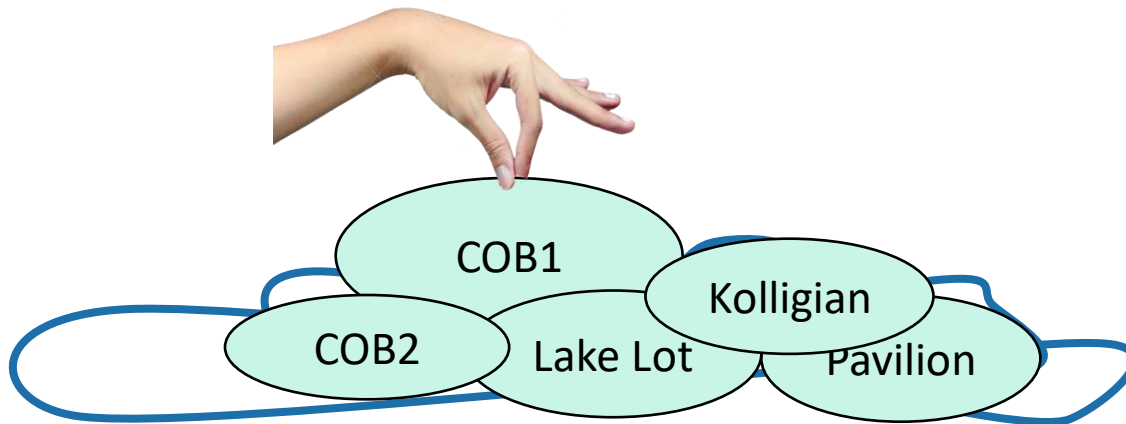
Dijkstra's algorithm

- Finds shortest paths from COB1 to everywhere else.



Dijkstra intuition

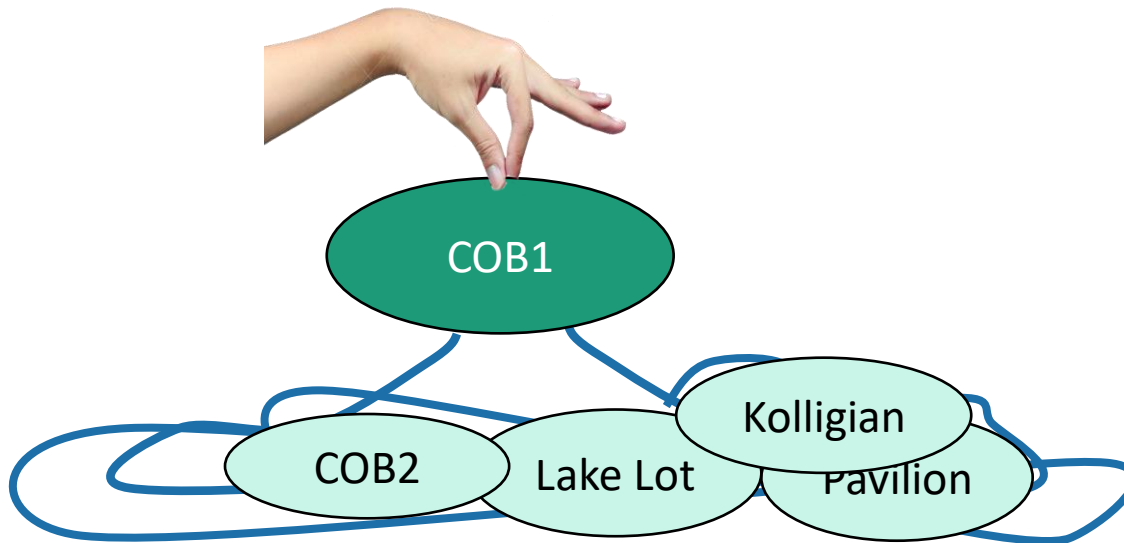
YOINK!



Dijkstra intuition

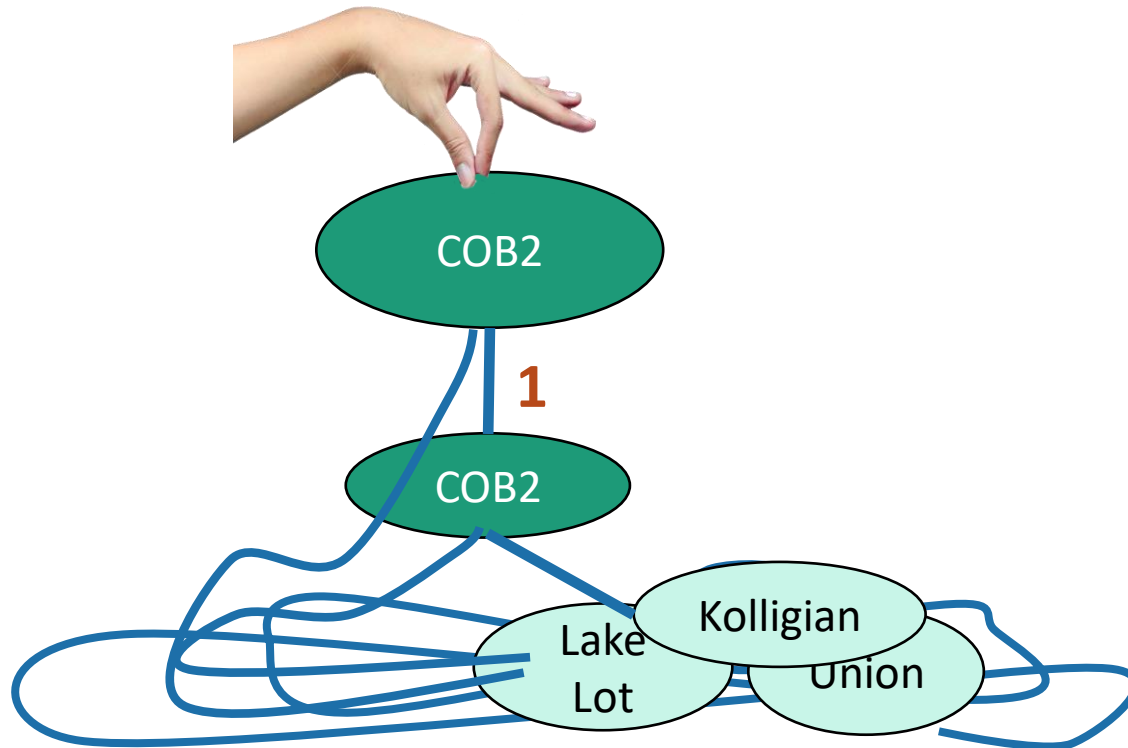
A vertex is done when it's not on the ground anymore.

YOINK!



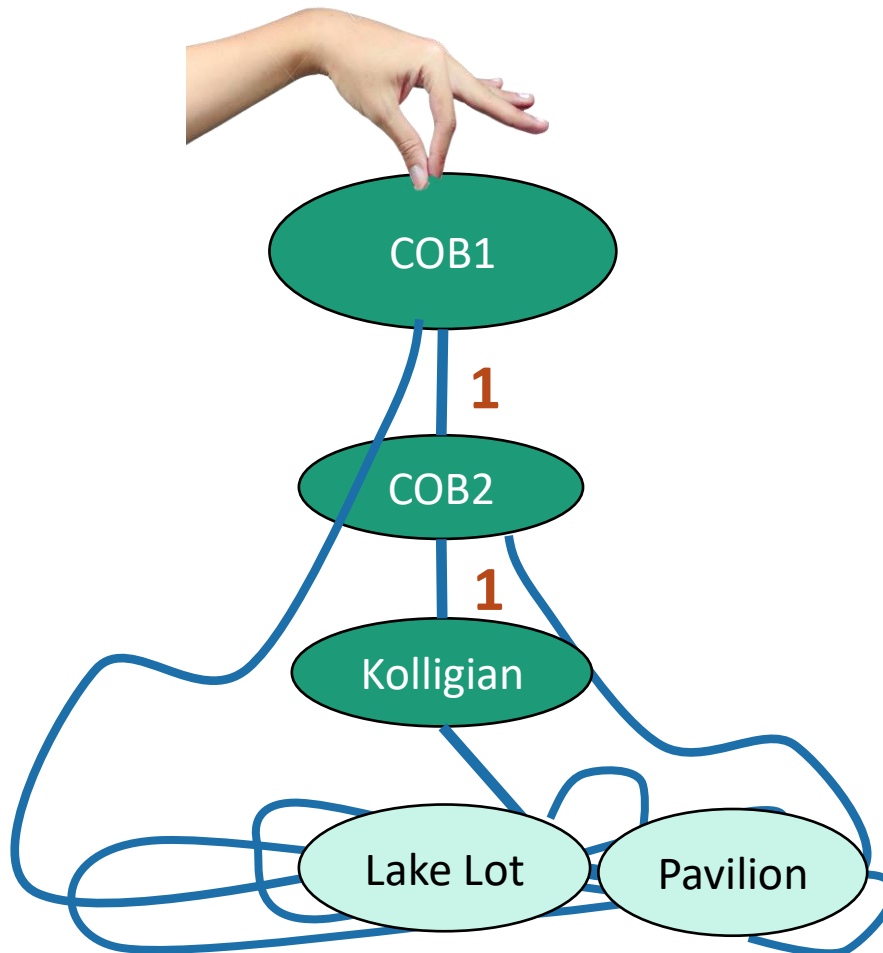
Dijkstra intuition

YOINK!



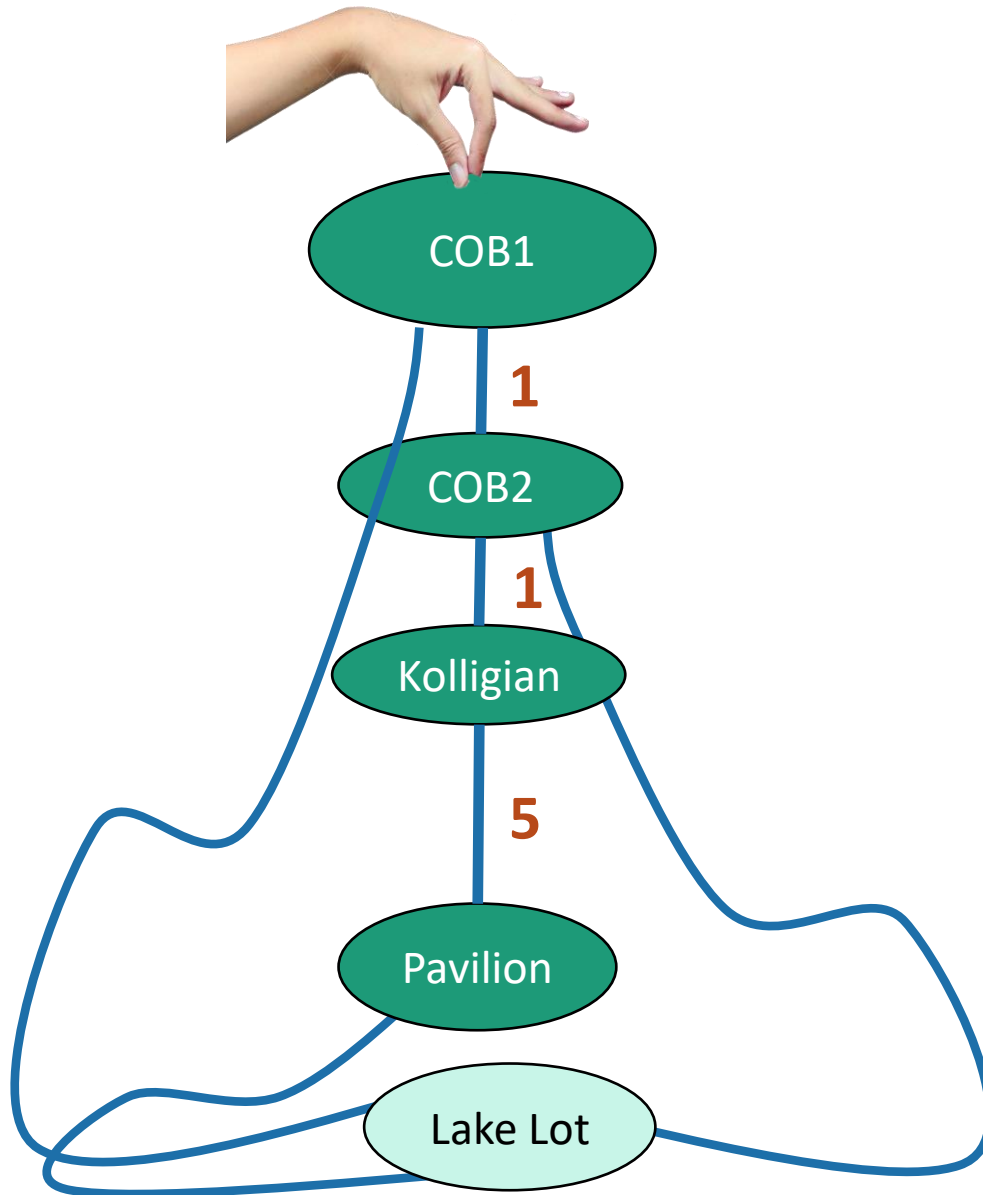
Dijkstra intuition

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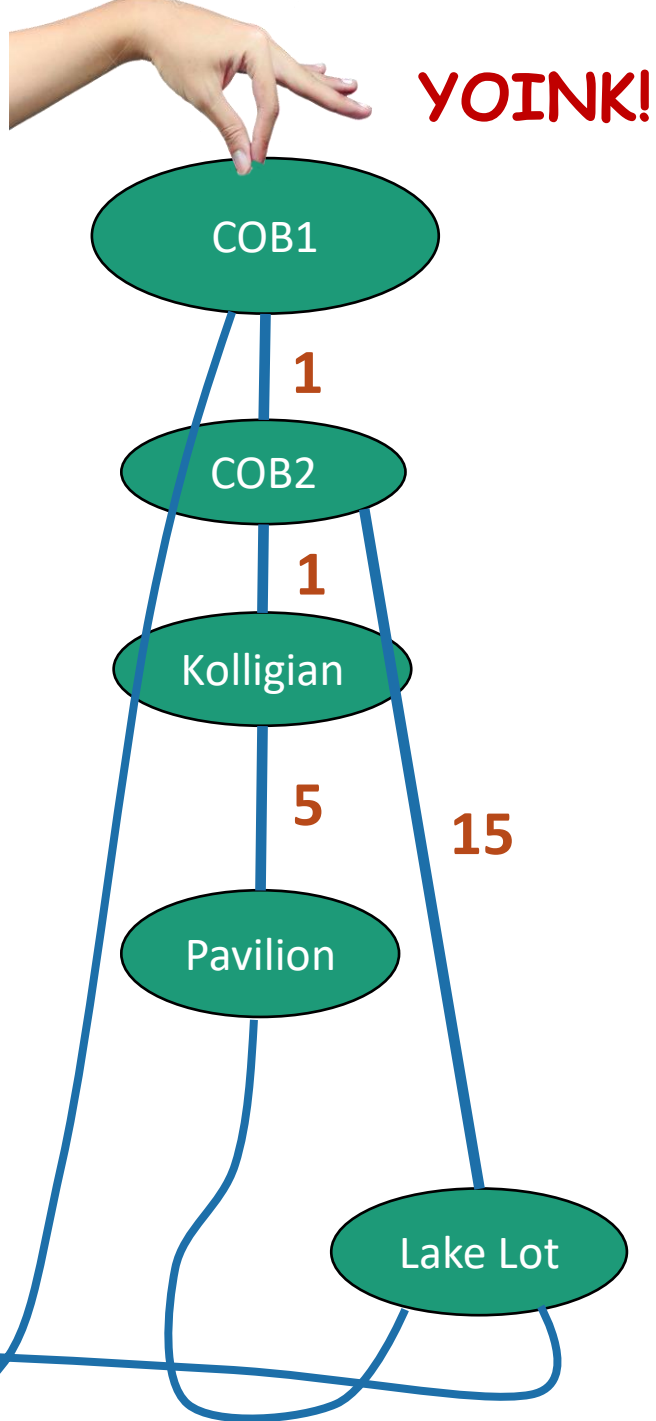


Dijkstra intuition

YOINK!



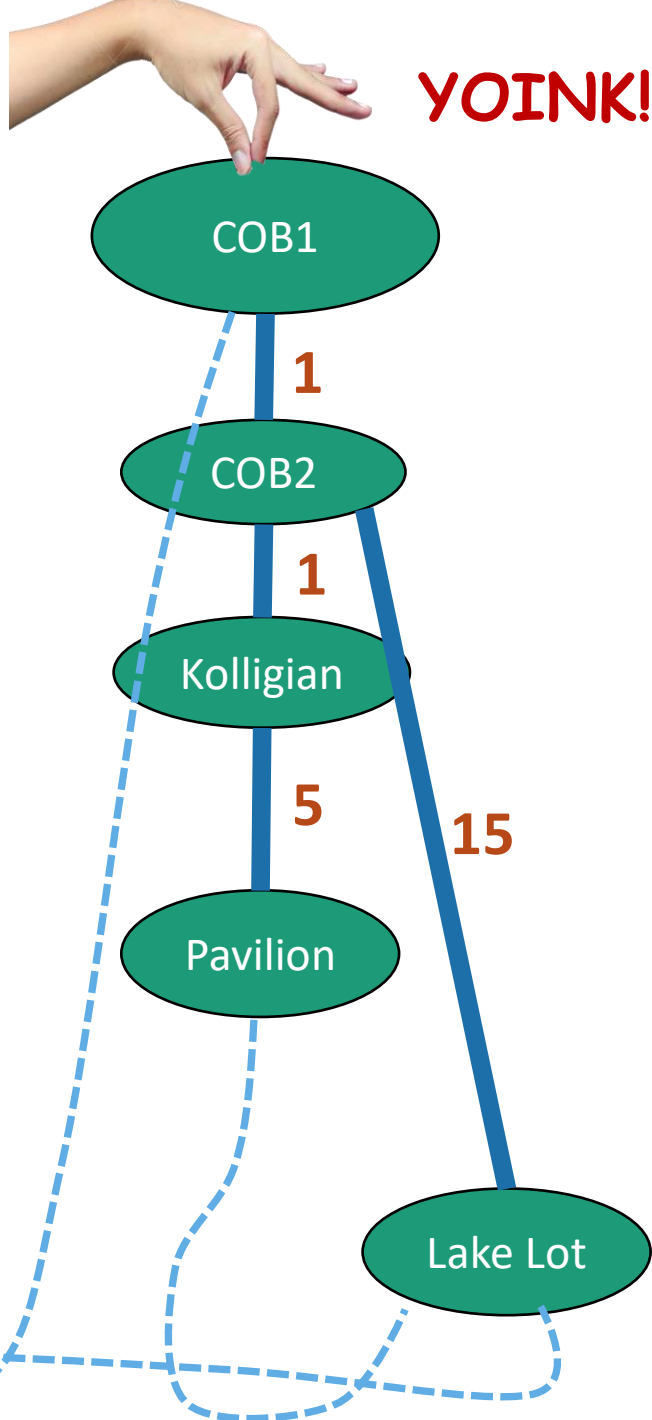
Dijkstra intuition



Dijkstra intuition

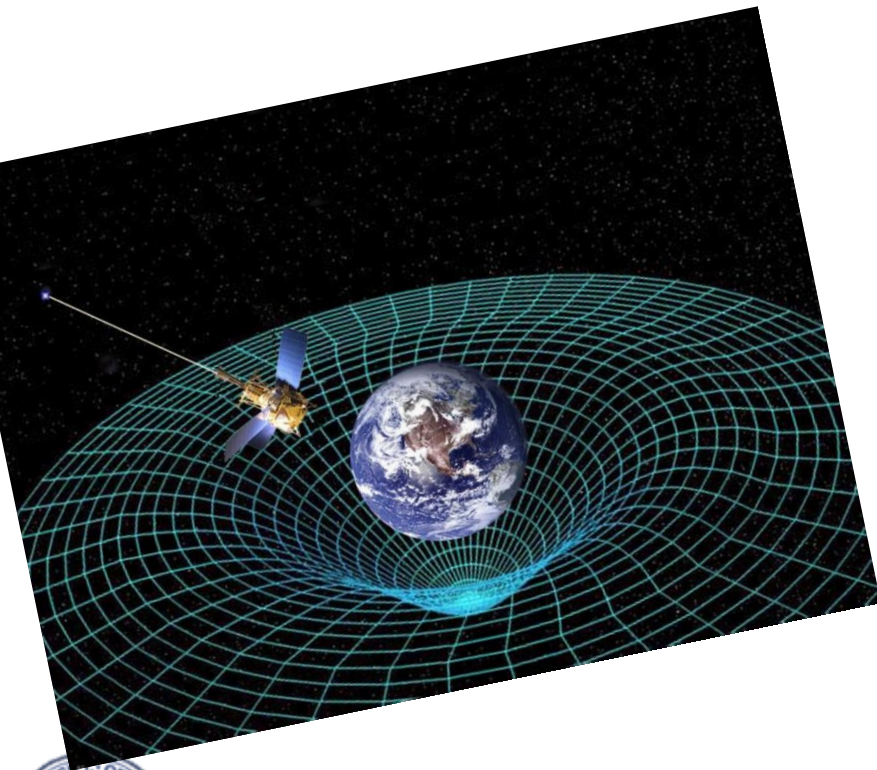
This creates a tree!

The shortest paths
are the lengths
along this tree.



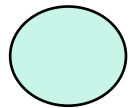
How do we actually implement this?

- **Without** string and gravity?

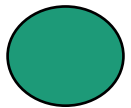


Dijkstra by example

How far is a node from COB1?



I'm not sure yet



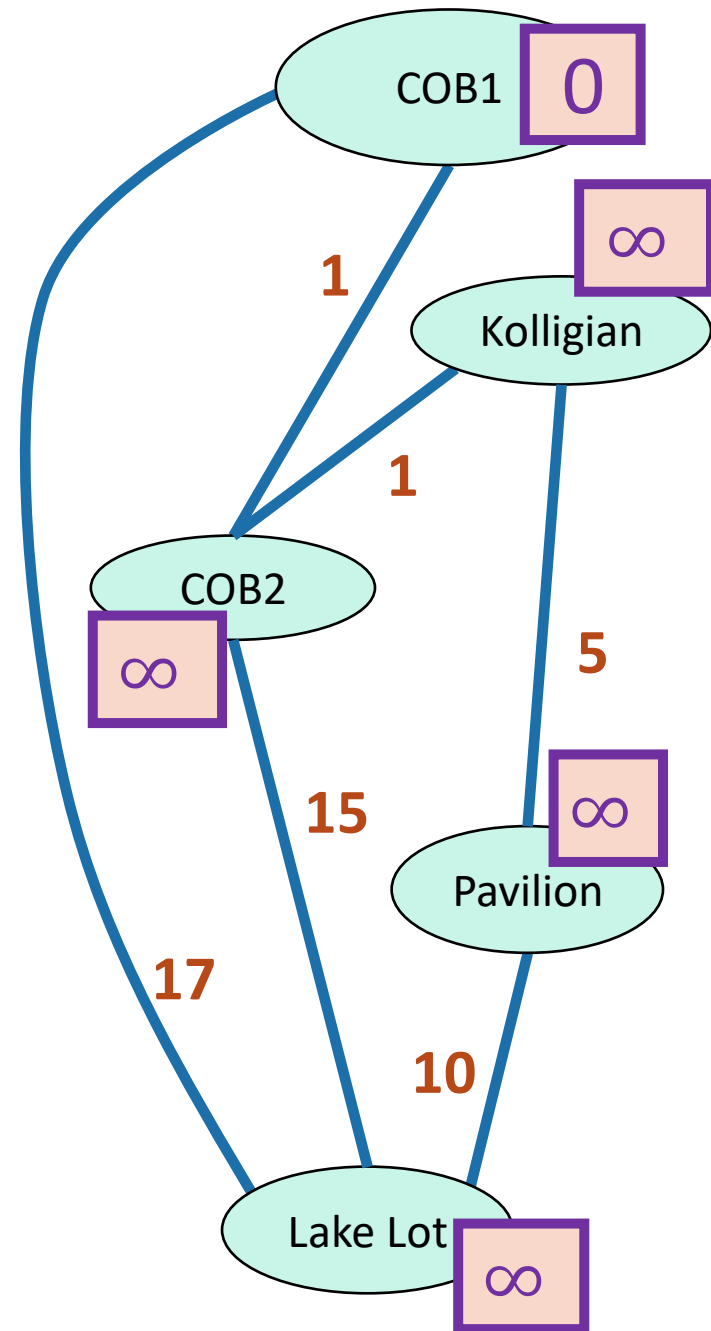
I'm sure



$x = d[v]$ is my best **over-estimate** for $\text{dist}(\text{COB1}, v)$.

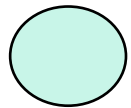
Initialize $d[v] = \infty$
for all non-starting vertices
 v , and $d[\text{COB1}] = 0$

- Pick the **not-sure** node u with the smallest estimate $d[u]$.

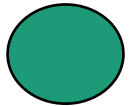


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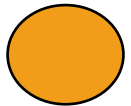
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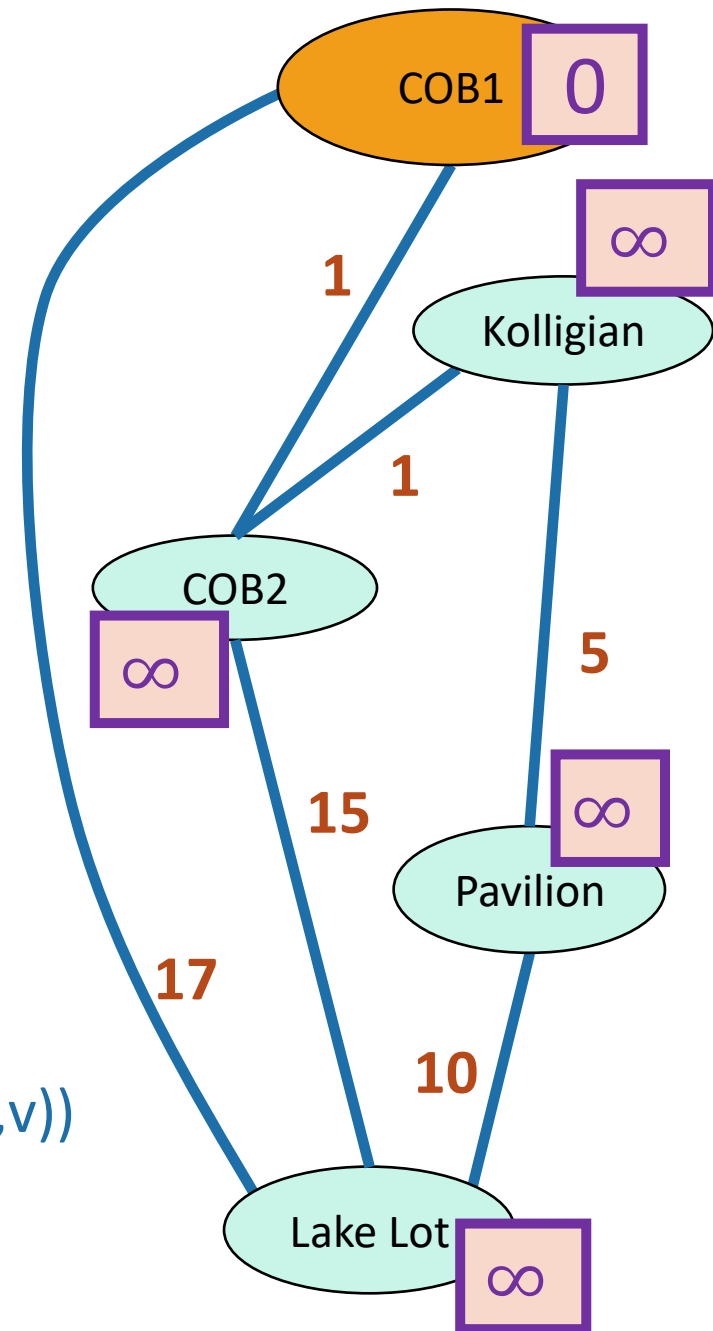


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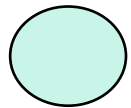
Current node u

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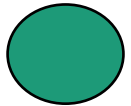


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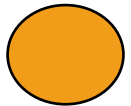
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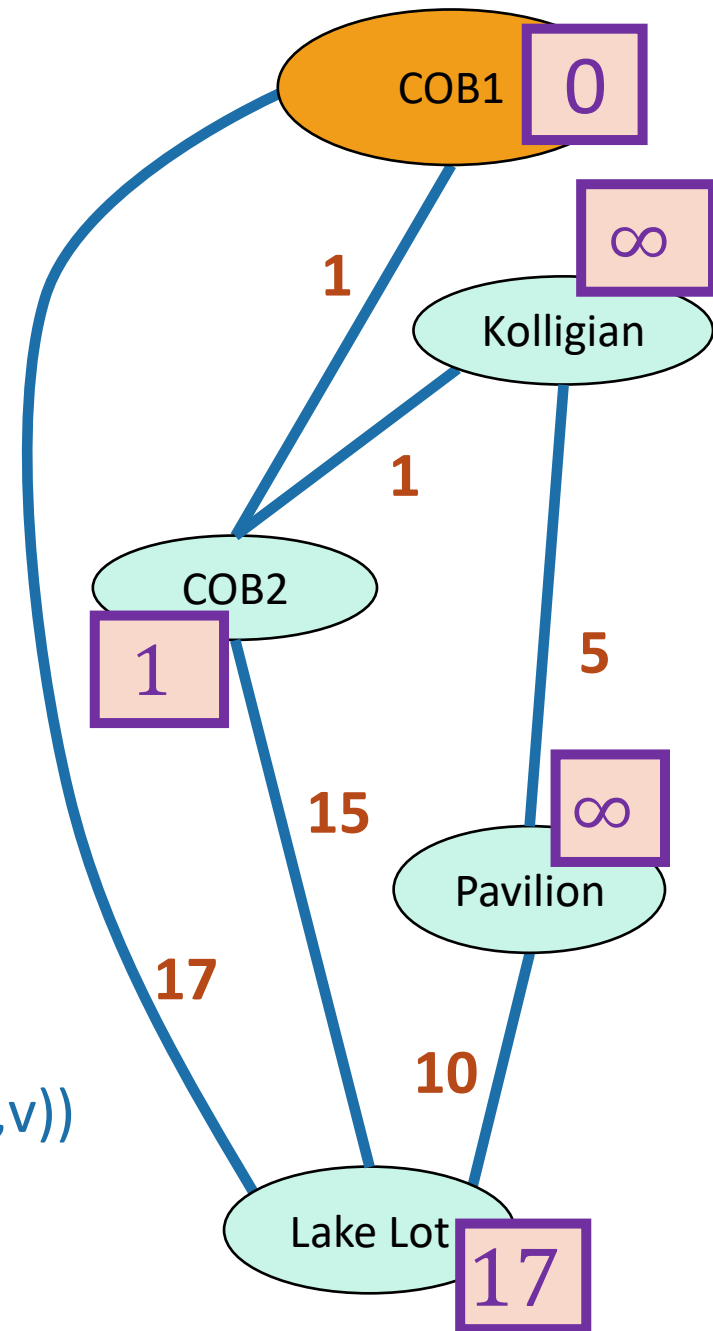


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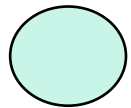
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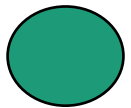


Dijkstra by example

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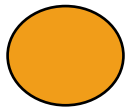
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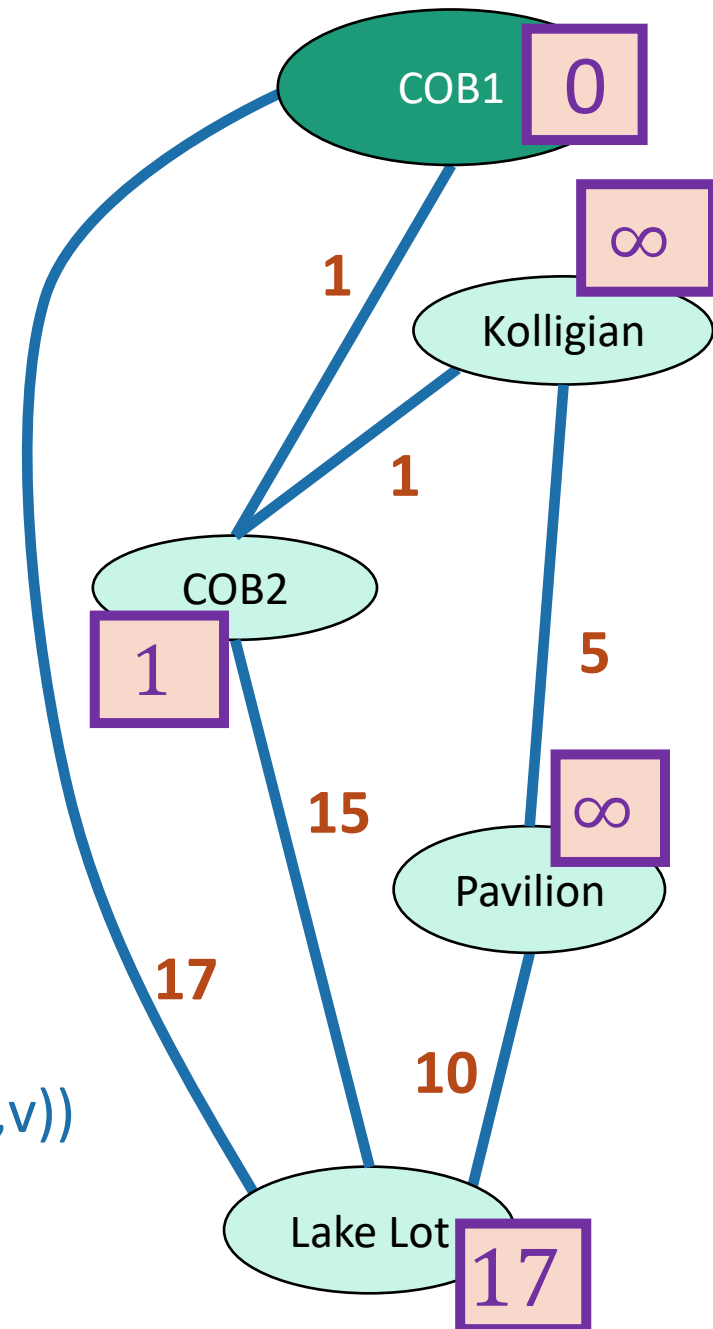


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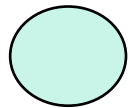
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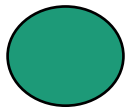


Dijkstra by example

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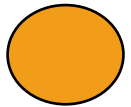
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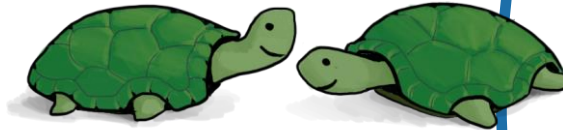
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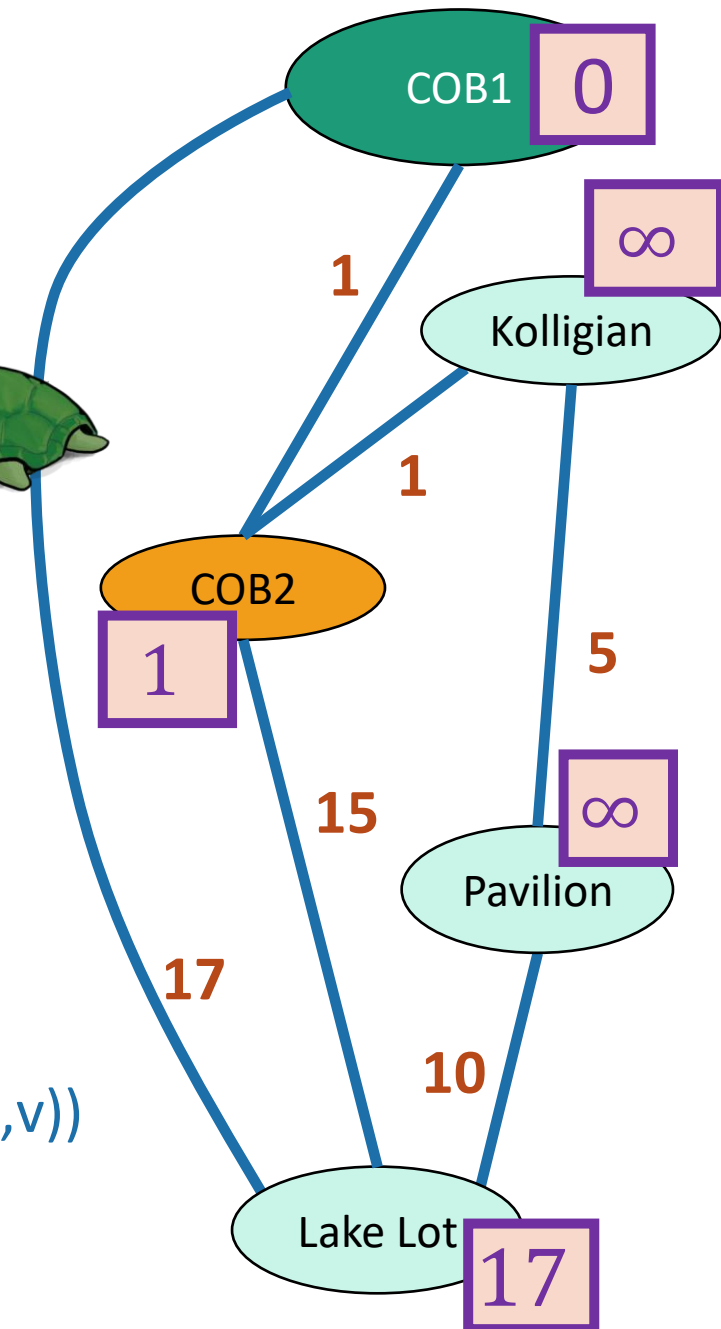


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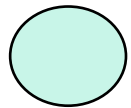
COB2 has three neighbors. What happens when we update them?

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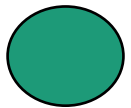


Dijkstra by example

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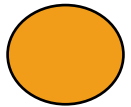
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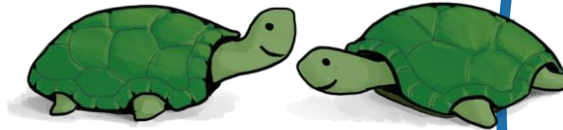
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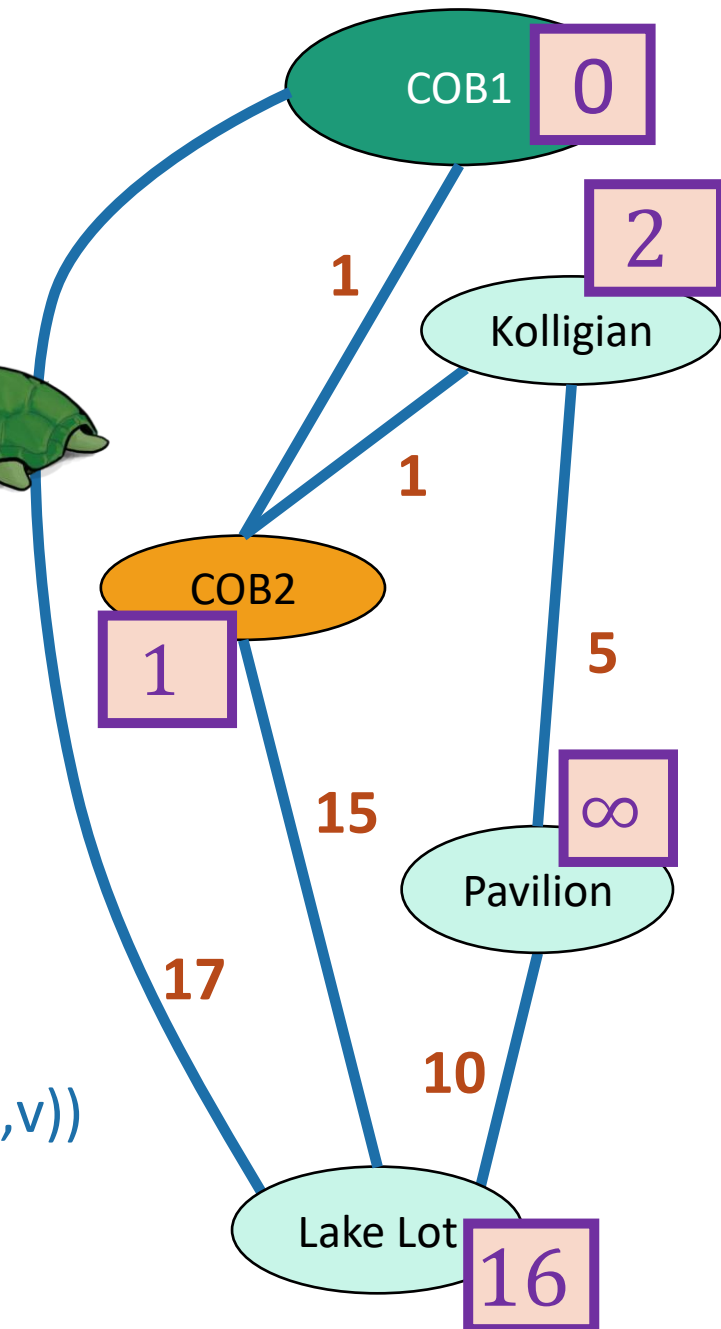


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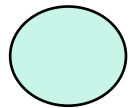
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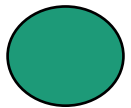


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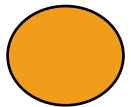
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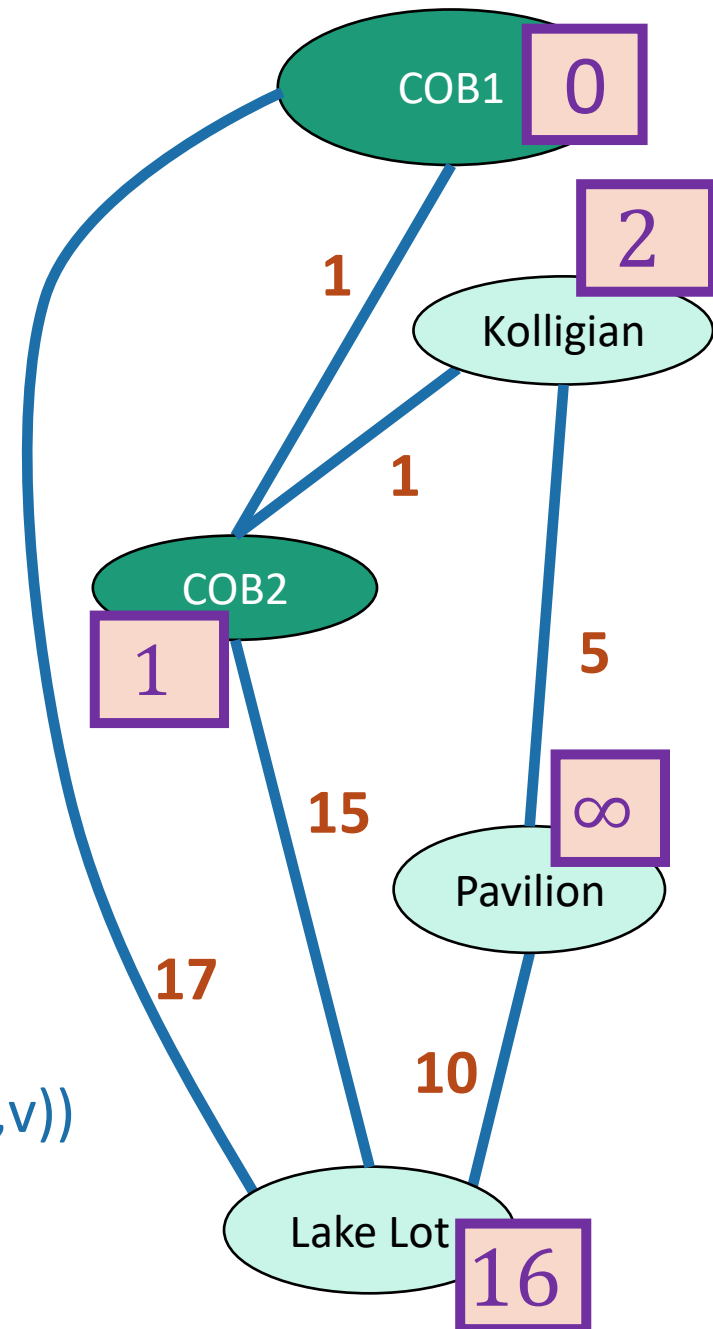


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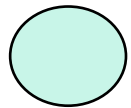
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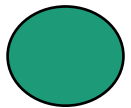


Dijkstra by example

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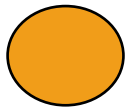
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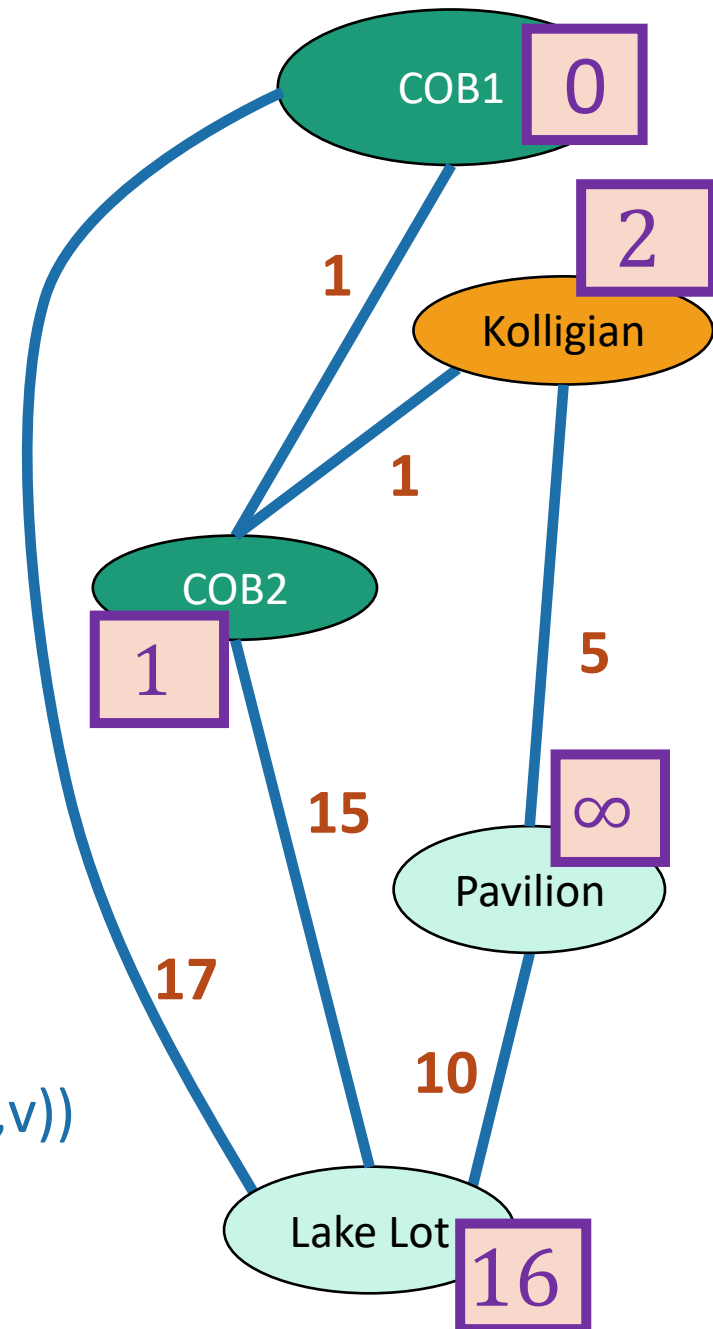


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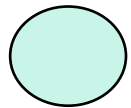
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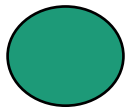


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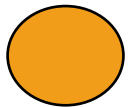
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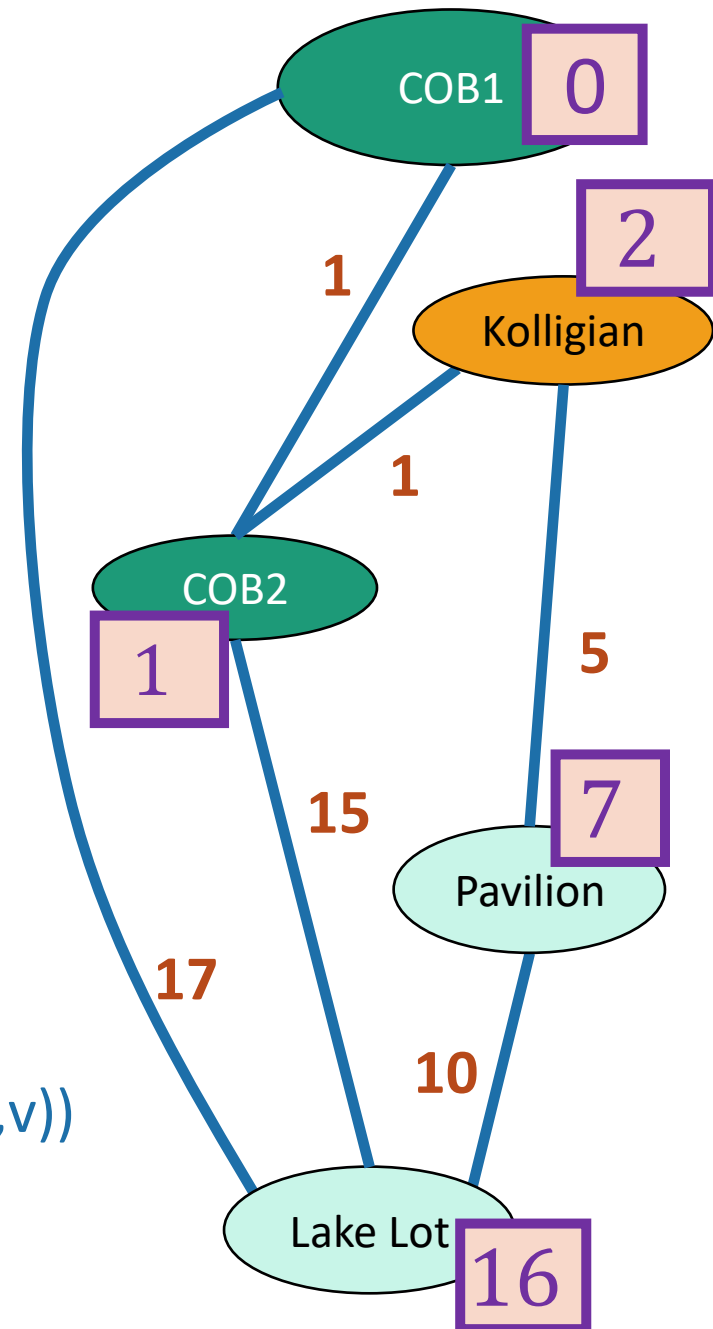


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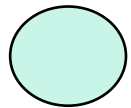
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- Pick the **not-sure** node u with the smallest estimate $d[u]$.
- Update all u 's neighbors v :
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- Mark u as **sure**.
- Repeat

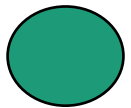


Dijkstra by example

How far is a node from COB1?



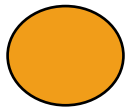
I'm not sure yet



I'm sure

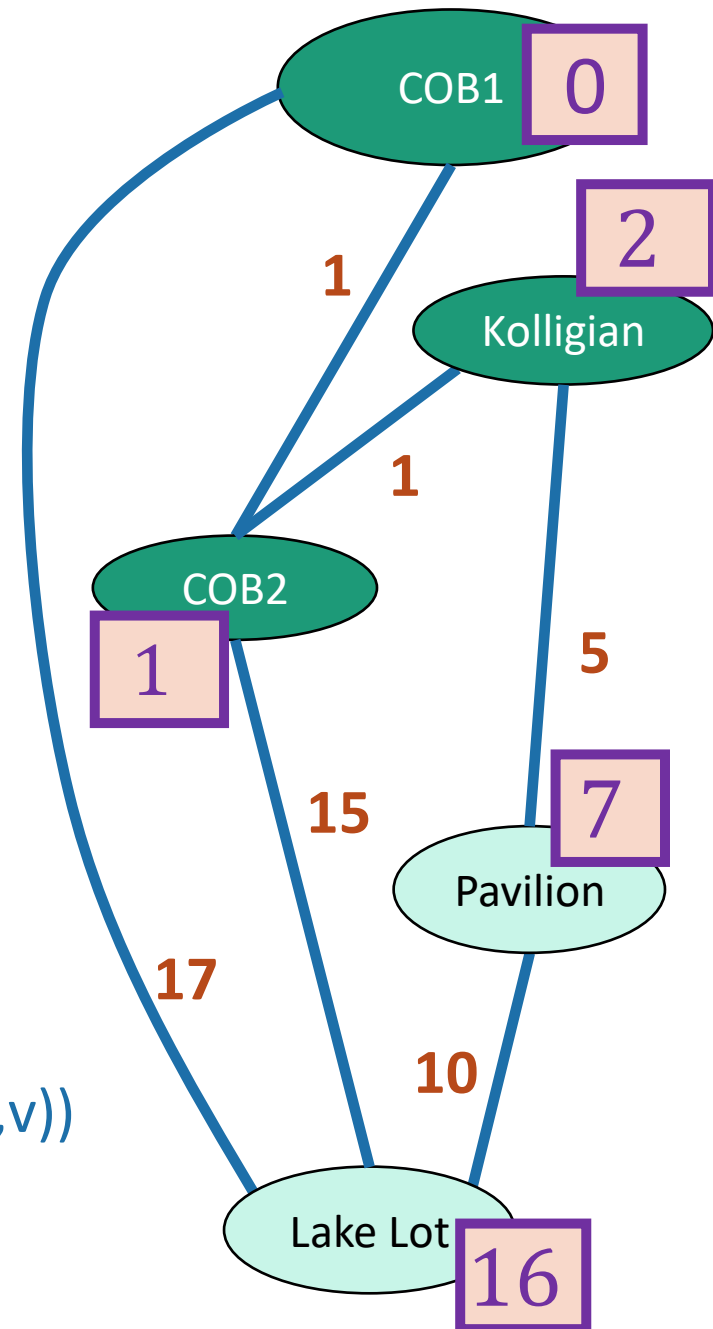


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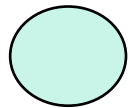
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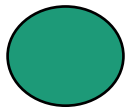


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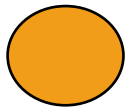
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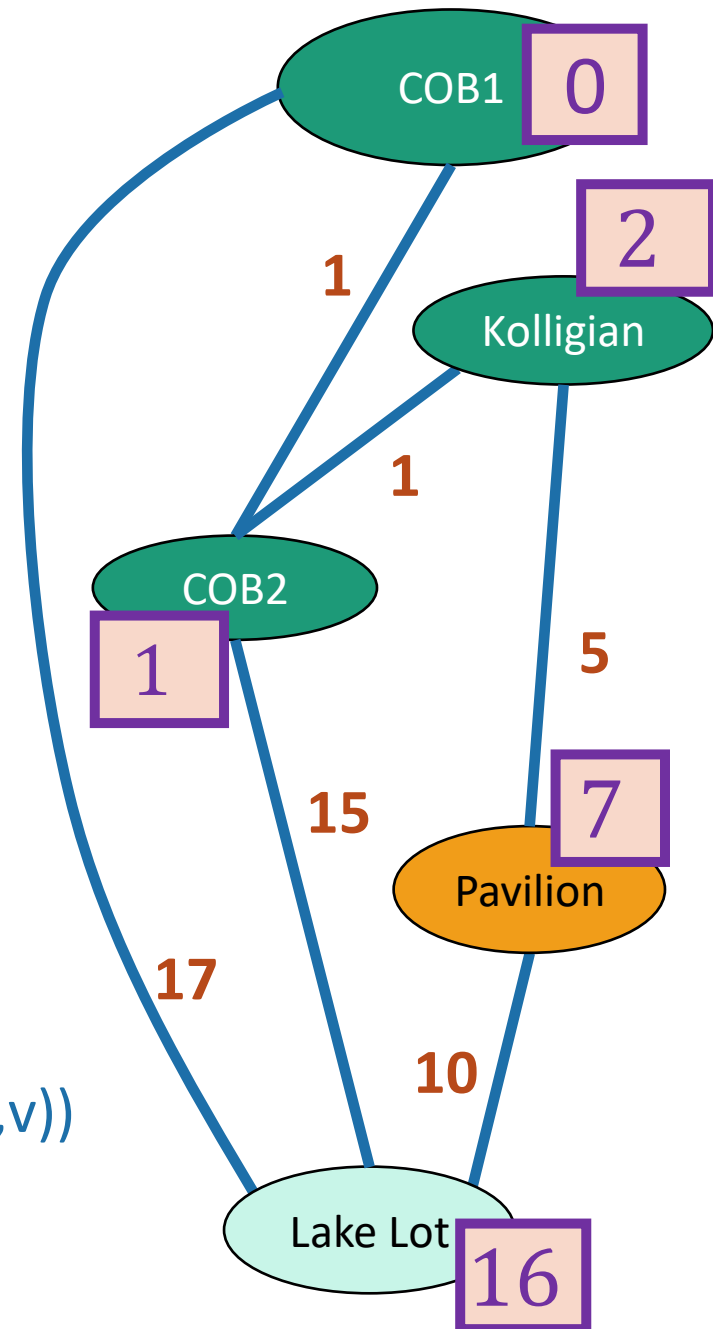


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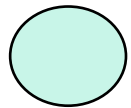
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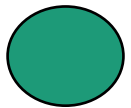


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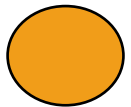
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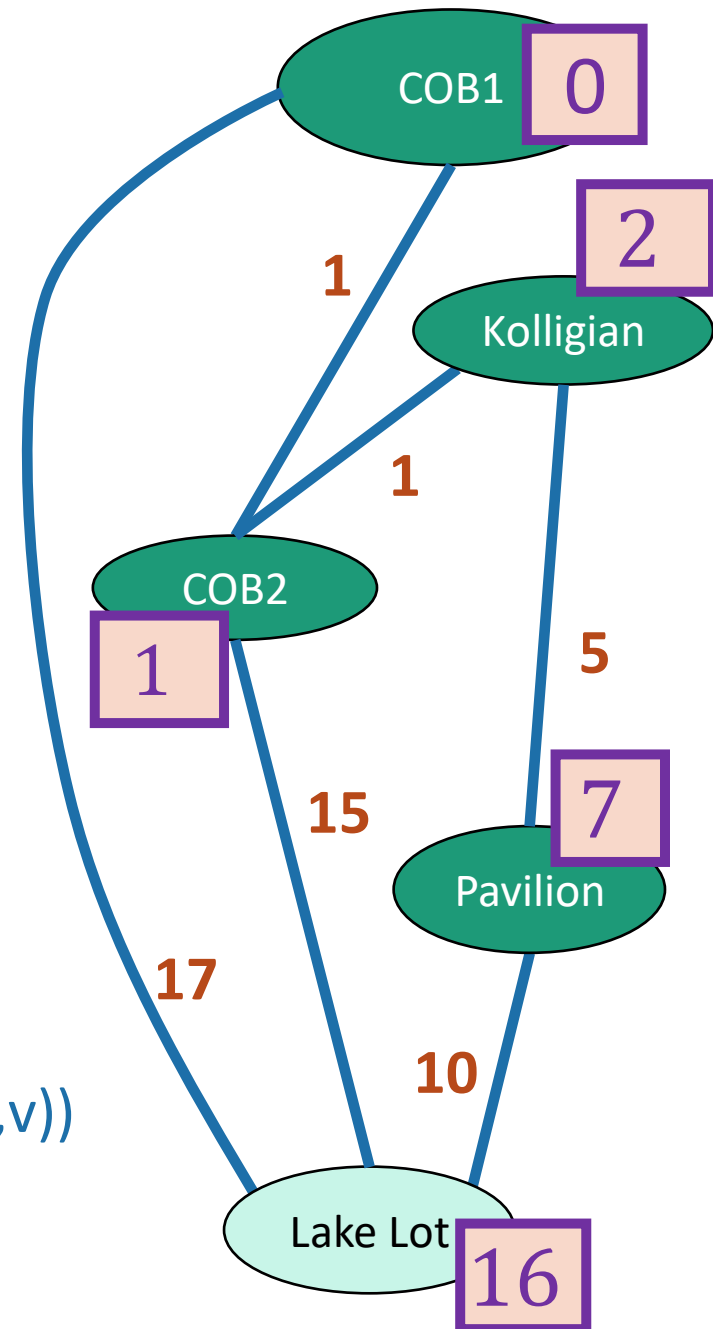


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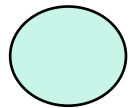
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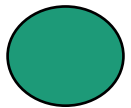


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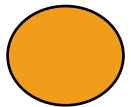
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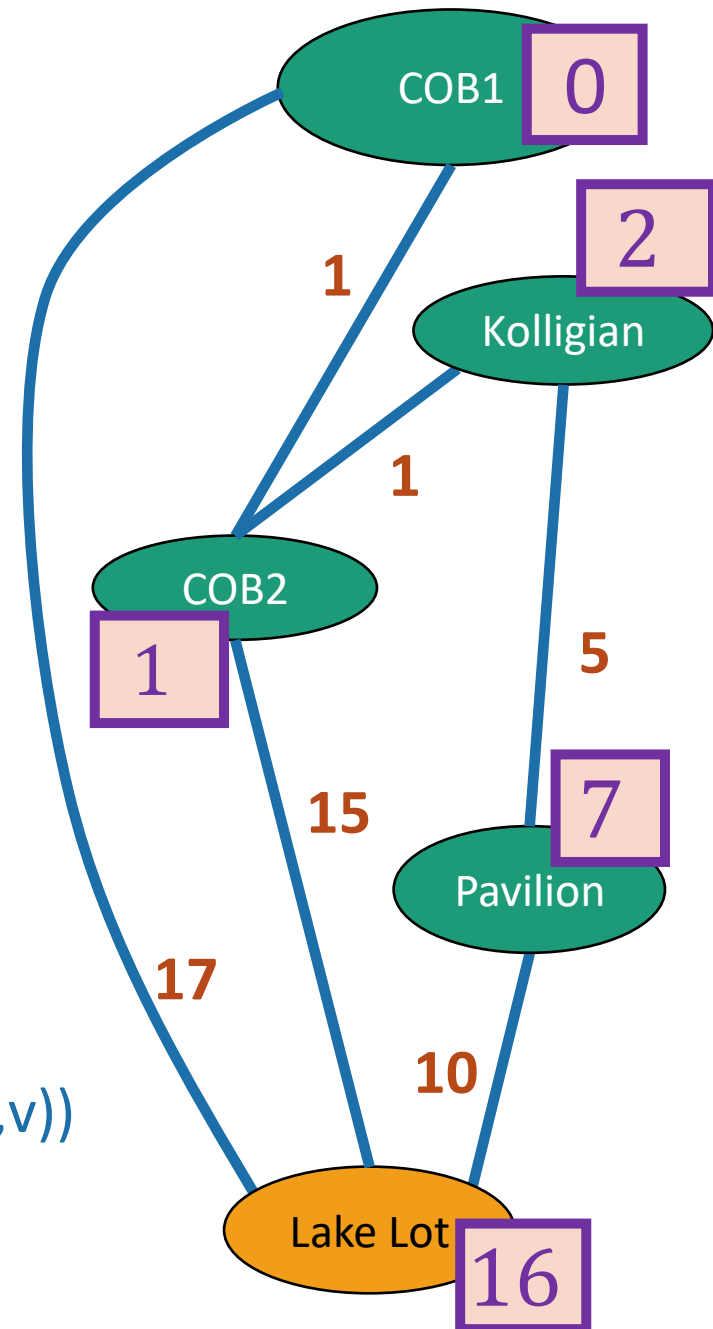


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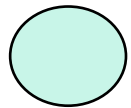
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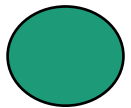


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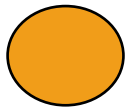
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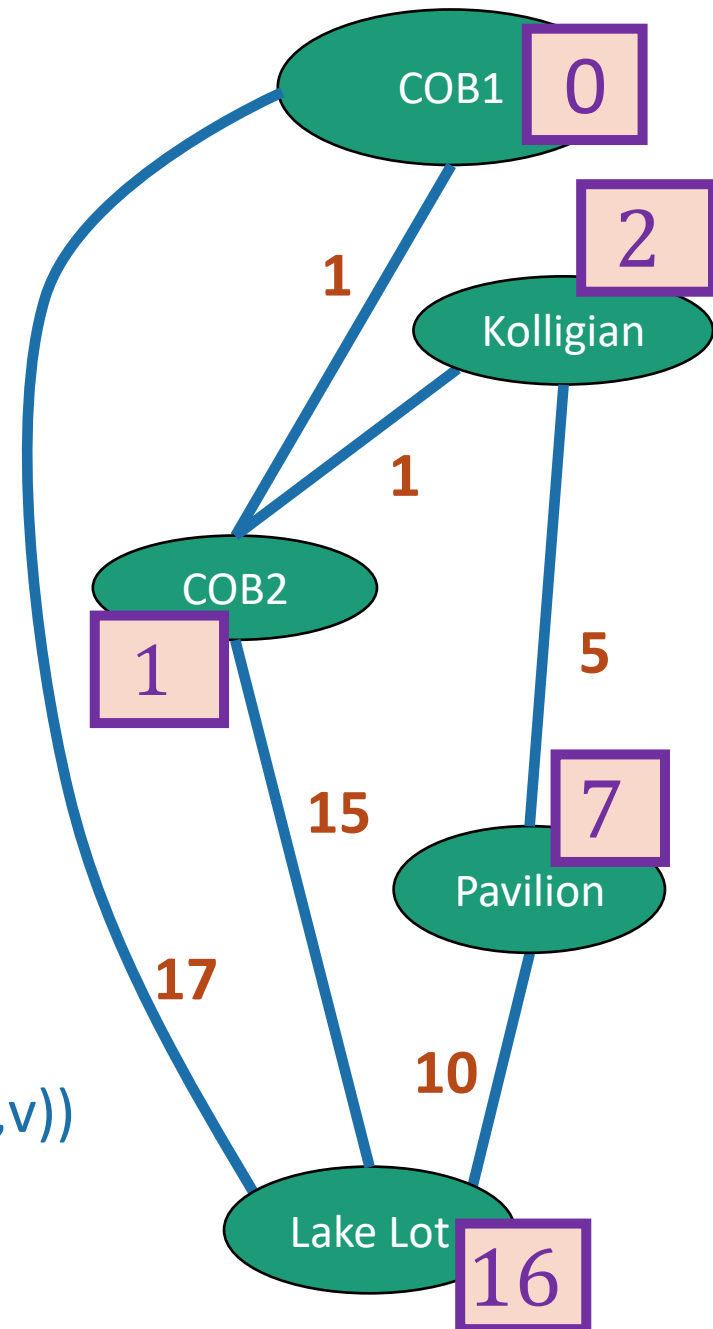


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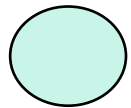
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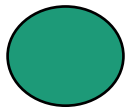


Dijkstra by example

How far is a node from COB1?



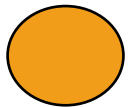
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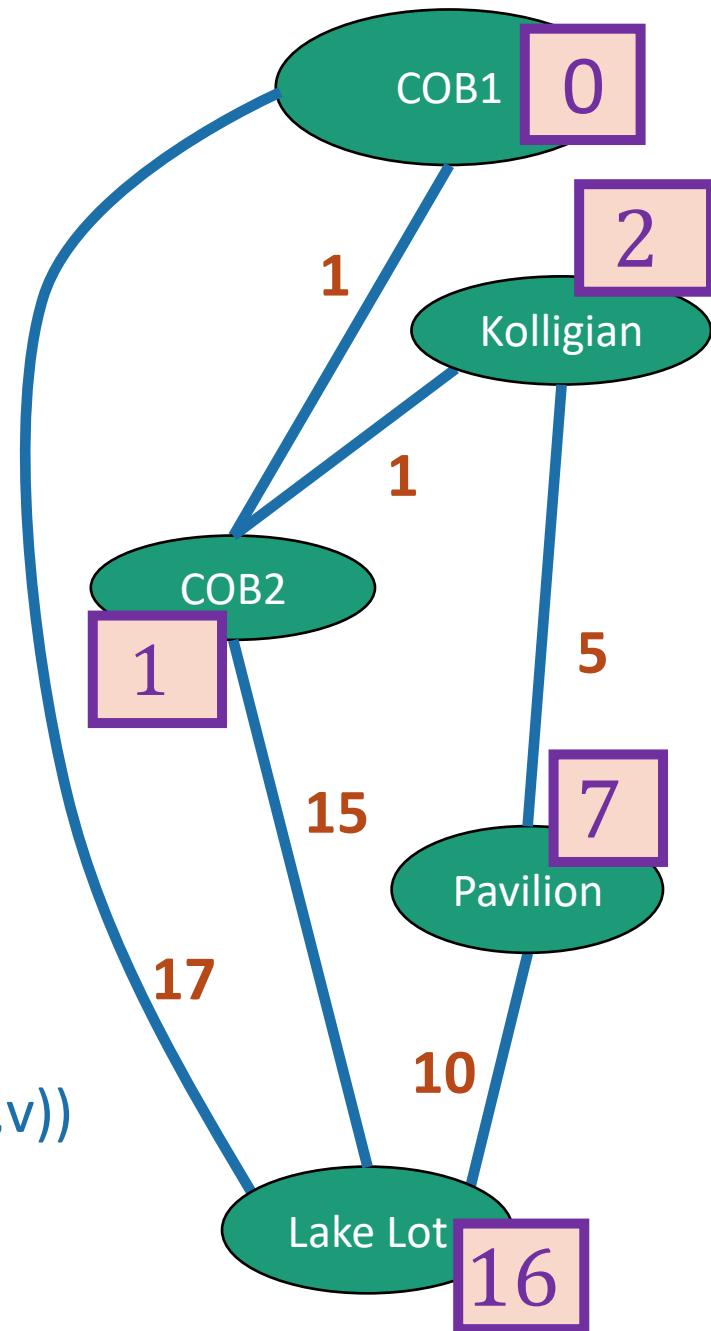


$x = d[v]$ is my best **over-estimate** for $\text{dist}(\text{COB1}, v)$.



Current node u

- Pick the **not-sure** node u with the smallest estimate $d[u]$.
- Update all u 's neighbors v :
 - $d[v] = \min(d[v], d[u] + \text{edgeWeight}(u, v))$
- Mark u as **sure**.
- Repeat
- After all nodes are **sure**, say that $d(\text{COB1}, v) = d[v]$ for all v



Dijkstra's algorithm

Dijkstra(G,s):

- Set all vertices to **not-sure**
- $d[v] = \infty$ for all v in V
- $d[s] = 0$
- **While** there are **not-sure** nodes:
 - Pick the **not-sure** node u with the smallest estimate **$d[u]$** .
 - **For** v in u .neighbors:
 - $d[v] \leftarrow \min(d[v] , d[u] + \text{edgeWeight}(u,v))$
 - Mark u as **sure**.
- Now $d(s, v) = d[v]$

Lots of implementation details left un-explained.
We'll get to that!



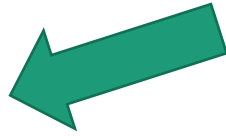
As usual

- Does it work?

- Yes.

- Is it fast?

- Depends on how you implement it.



Why does this work?

- **Theorem:**

- Suppose we run Dijkstra on $G=(V,E)$, starting from s .
- At the end of the algorithm, the estimate $d[v]$ is the actual distance $d(s,v)$.

Let's rename "COB1" to "s", our starting vertex.

- **Proof outline:**

- **Claim 1:** For all v , $d[v] \geq d(s,v)$.
- **Claim 2:** When a vertex v is marked **sure**, $d[v] = d(s,v)$.

- **Claims 1 and 2** imply the **theorem**.

- When v is marked **sure**, $d[v] = d(s,v)$. ← Claim 2
- $d[v] \geq d(s,v)$ and never increases, so after v is **sure**, $d[v]$ stops changing. ← Claim 1 + def of algorithm
- This implies that at any time *after* v is marked **sure**, $d[v] = d(s,v)$.
- All vertices are **sure** at the end, so all vertices end up with $d[v] = d(s,v)$.

Next let's prove the claims!



Claim 1

$d[v] \geq d(s,v)$ for all v .

Informally:

- Every time we update $d[v]$, we have a path in mind:

$$d[v] \leftarrow \min(\text{d[v]}, \text{d[u]} + \text{edgeWeight(u,v)})$$

Whatever path we
had in mind before

The shortest path to u , and
then the edge from u to v .

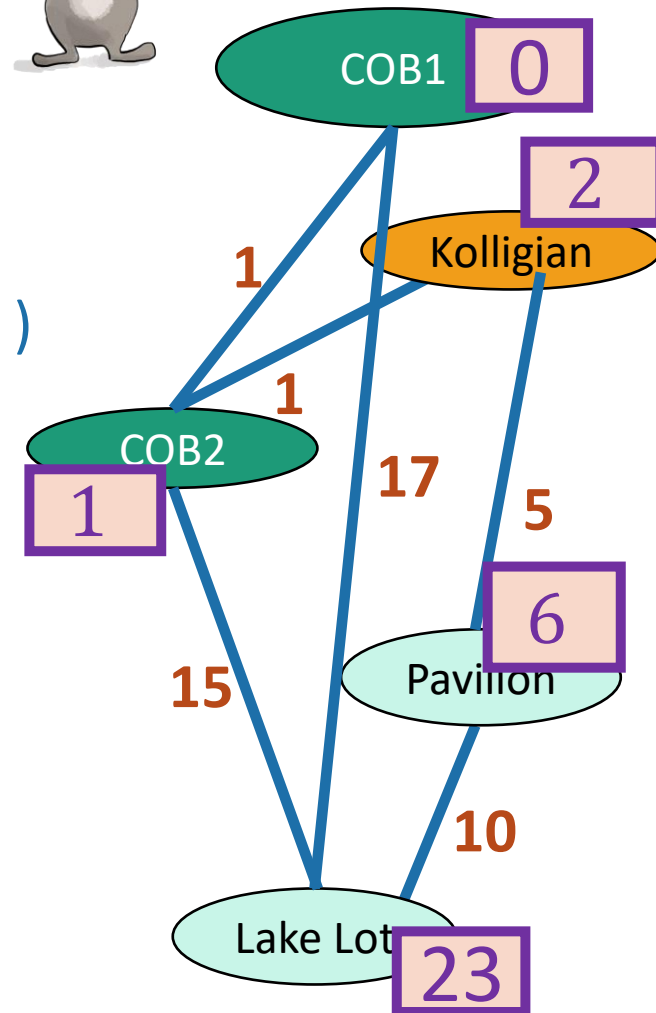
- $d[v]$ = length of the path we have in mind
 \geq length of shortest path
 $= d(s,v)$

Formally:

- We should prove this by induction.
- (See next slide or do it yourself)



Intuition!



Claim 1

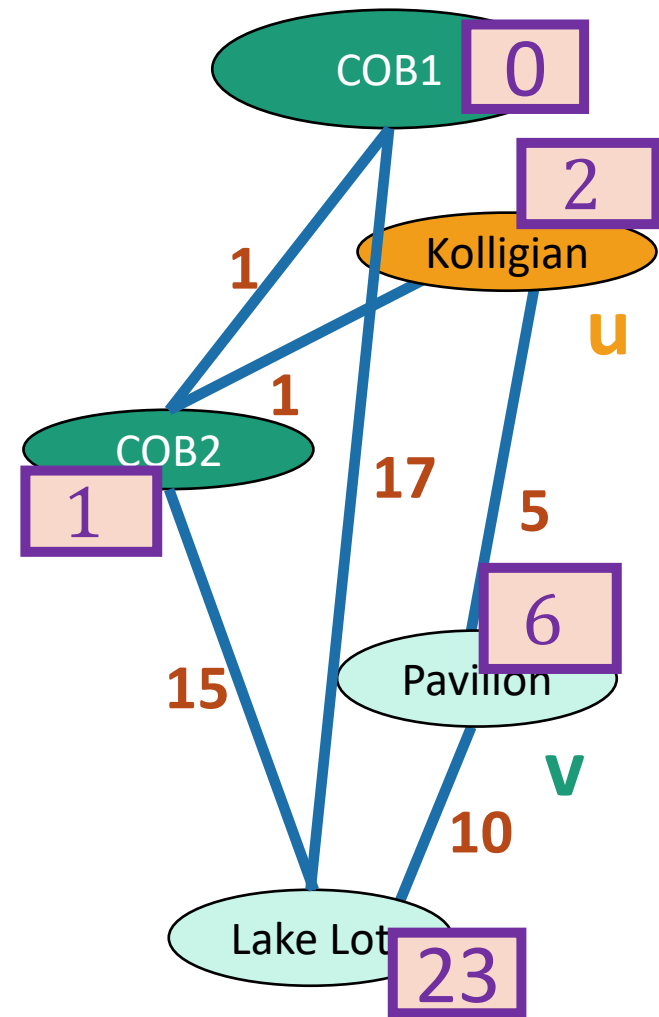
$d[v] \geq d(s,v)$ for all v .

- Inductive hypothesis.
 - After t iterations of Dijkstra,
 $d[v] \geq d(s,v)$ for all v .
- Base case:
 - At step 0, $d(s,s) = 0$, and $d(s,v) \leq \infty$
- Inductive step: say hypothesis holds for t .
 - At step $t+1$:
 - Pick u ; for each neighbor v :
 - $d[v] \leftarrow \min(d[v], d[u] + w(u,v))$

By induction,
 $d(s,v) \leq d[v]$

$d(s,v) \leq d(s,u) + d(u,v)$
 $\leq d[u] + w(u,v)$
using induction again for $d[u]$

So the inductive
hypothesis holds
for $t+1$, and Claim
1 follows.



Claim 2

When a vertex u is marked sure, $d[u] = d(s,u)$

- Inductive Hypothesis:

- When we mark the t' 'th vertex v as **sure**, $d[v] = d(s,v)$.

- Base case:

- The first vertex marked **sure** is s , and $d[s] = d(s,s) = 0$.

(Note: we are assuming here that the edge weights are non-negative, so there's no way to sneakily get from s to s with cost less than zero!)

- Inductive step:

- Suppose that we are about to add u to the **sure** list.
- That is, we picked u in the first line here:

- Pick the **not-sure** node u with the smallest estimate $d[u]$.
- Update all u 's neighbors v :
 - $d[v] \leftarrow \min(d[v], d[u] + \text{edgeWeight}(u,v))$
- Mark u as **sure**.
- Repeat

- Assume by induction that every v already marked **sure** has $d[v] = d(s,v)$.
- Want to show that $d[u] = d(s,u)$.



YOINK!

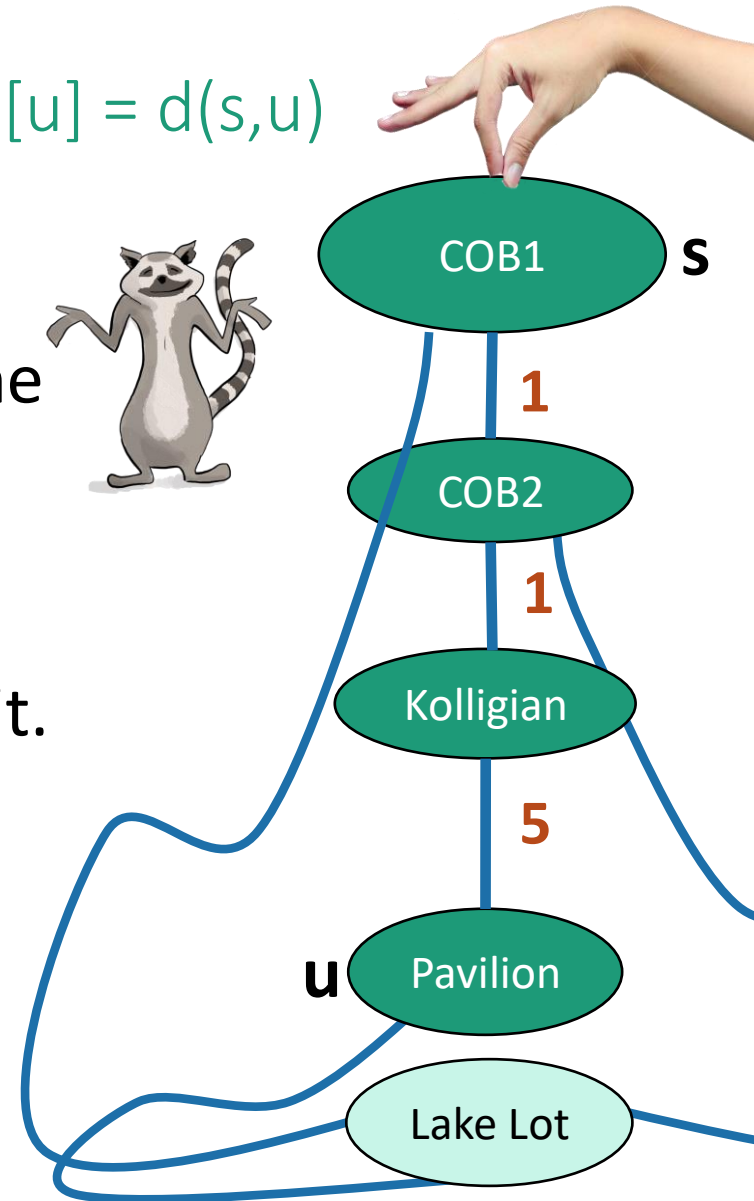
Intuition

When a vertex u is marked sure, $d[u] = d(s, u)$

- The first path that lifts u off the ground is the shortest one.



- But we should actually prove it.



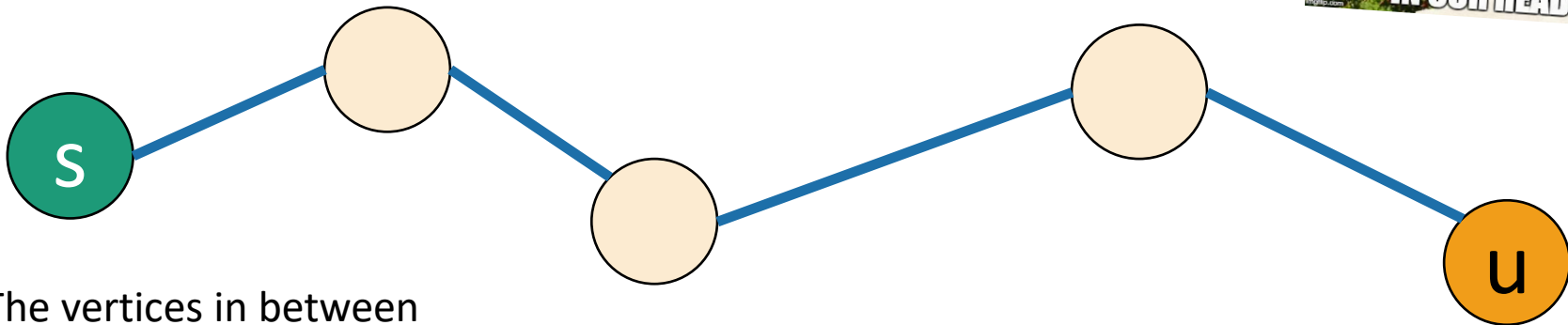
Claim 2

Inductive step

Temporary definition:

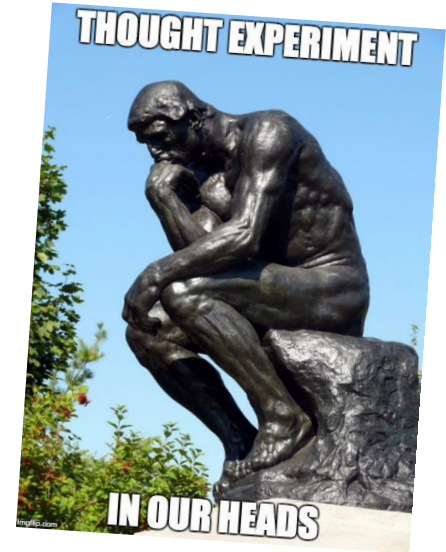
v is “good” means that $d[v] = d(s, v)$

- Want to show that u is good.
- Consider a **true** shortest path from s to u :



The vertices in between are beige because they may or may not be **sure**.

True shortest path.



Claim 2

Inductive step

Temporary definition:

v is “good” means that $d[v] = d(s, v)$



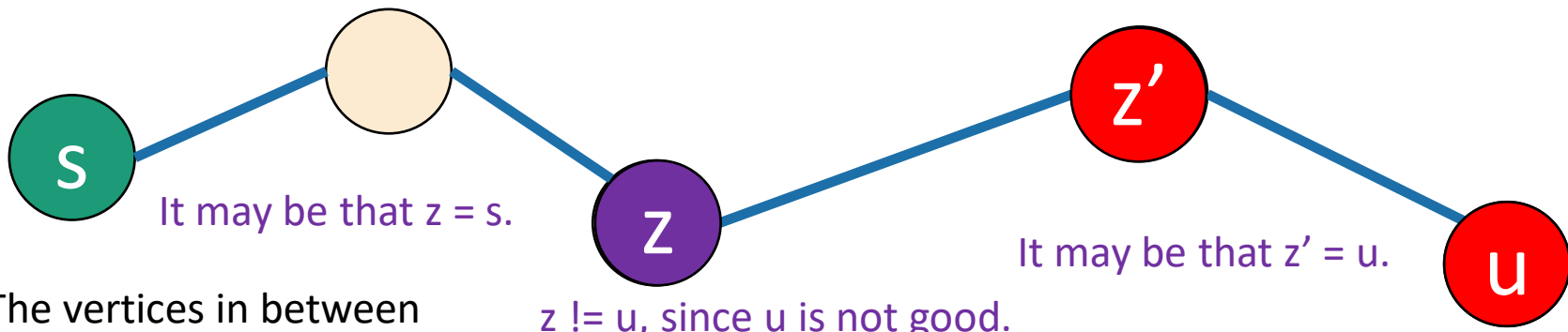
means good



means not good

“by way of contradiction”

- Want to show that u is good. BWOC, suppose u isn't good.
- Say z is the last good vertex before u .
- z' is the vertex after z .



The vertices in between are beige because they may or may not be **sure**.

True shortest path.

Claim 2

Inductive step

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v is “good” means that $d[v] = d(s, v)$



means good



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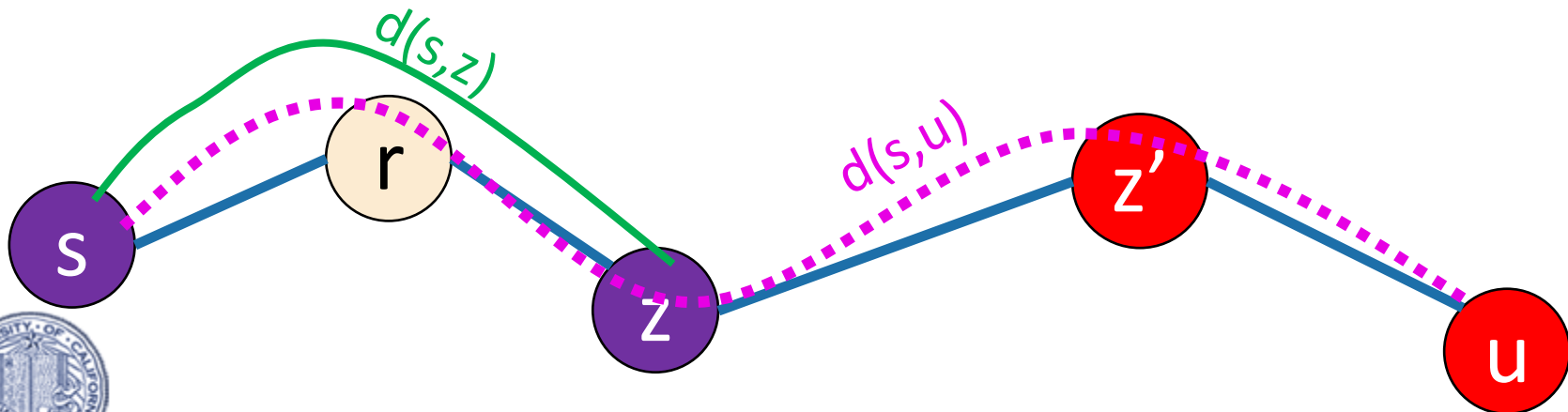
- Want to show that u is good. BWOC, suppose u isn't good.

$$d[z] = d(s, z) \leq d(s, u) \leq d[u]$$

z is good

Subpaths of
shortest paths are
shortest paths.

AND, also that $d(z, u) \geq 0$,
since all of the edge-weights
are non-negative!



Claim 2

Inductive step

Temporary definition:

v is “good” means that $d[v] = d(s, v)$



means good



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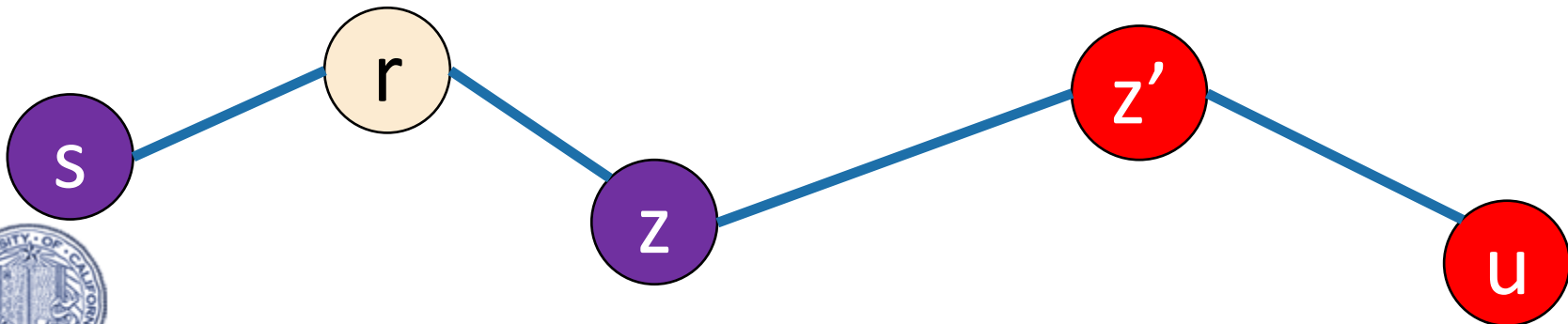
z is good

Subpaths of
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Claim 1

- If $d[z] = d[u]$, then u is good. ⚡ But u is not good!
- So $d[z] < d[u]$, so z is **sure**.

We chose u so that $d[u]$ was
smallest of the unsure vertices.



Claim 2

Inductive step

Temporary definition:

v is “good” means that $d[v] = d(s, v)$



means good



means not good

- Want to show that u is good. BWOC, suppose u isn't good.

$$d[z] = d(s, z) \leq d(s, u) \leq d[u]$$

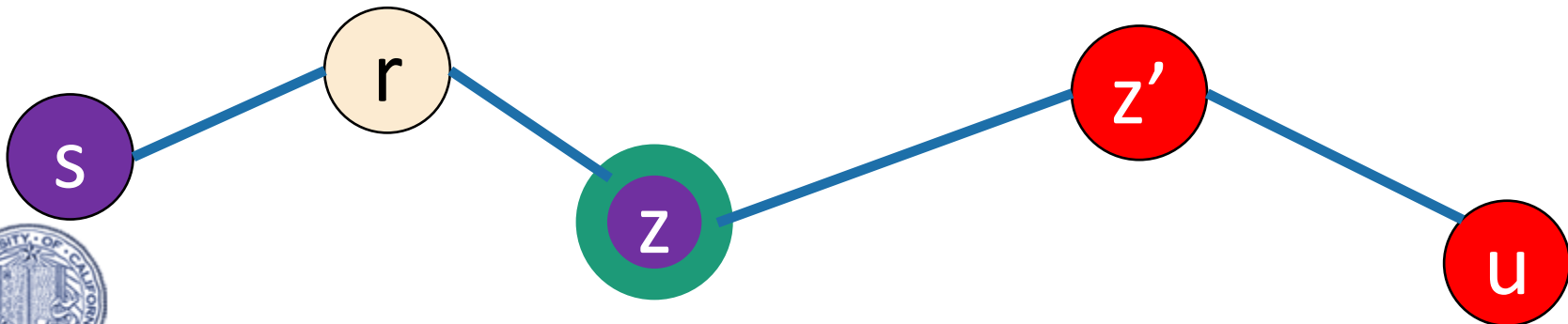
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Claim 2

Inductive step

Temporary definition:

v is “good” means that $d[v] = d(s, v)$



means good



means not good

• Want to show that u is good. BWOC, suppose u isn't good.

• If z is **sure** then we've already updated z' :

• $d[z'] \leq d[z] + w(z, z')$ **def of update**

$$d[z'] \leftarrow \min\{d[z'], d[z] + w(z, z')\}$$

$$= d(s, z) + w(z, z')$$

By induction when z was added to the sure list it had $d(s, z) = d[z]$

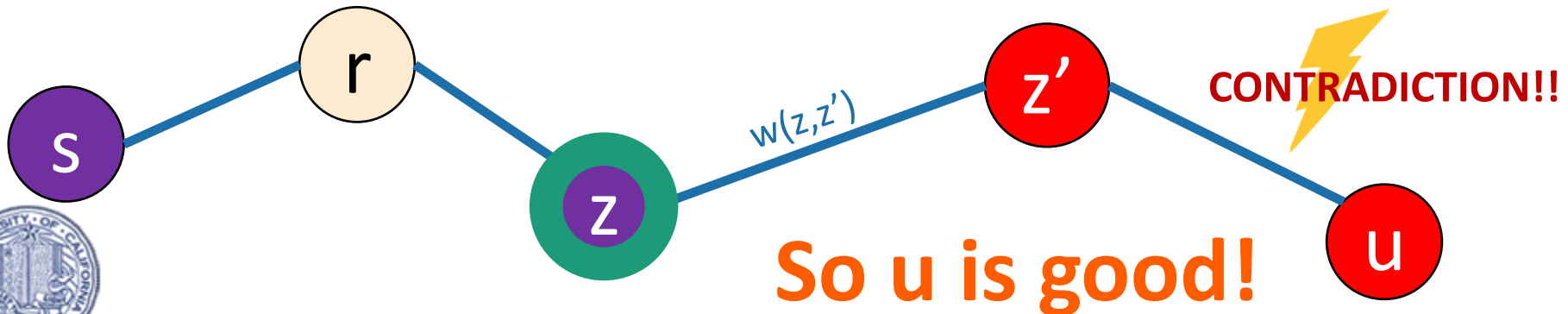
$$= d(s, z')$$

sub-paths of shortest paths are shortest paths

$$\leq d[z'] \text{ Claim 1}$$

So $d(s, z') = d[z']$ and so z' is good.

That is, the value of $d[z]$ when z was marked sure...



Claim 2

Back to this slide

When a vertex u is marked sure, $d[u] = d(s,u)$

- Inductive Hypothesis:

- When we mark the t' 'th vertex v as sure, $d[v] = \text{dist}(s,v)$.

- Base case:

- The first vertex marked **sure** is s , and $d[s] = d(s,s) = 0$.

- Inductive step:

- Suppose that we are about to add u to the **sure** list.
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- Repeat

- Assume by induction that every v already marked **sure** has $d[v] = d(s,v)$.
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Why does this work?

*Now back to
this slide*

- **Theorem:**

- Run Dijkstra on $G = (V, E)$ starting from s .
- At the end of the algorithm, the estimate $d[v]$ is the actual distance $d(s, v)$.

- Proof outline:

- **Claim 1:** For all v , $d[v] \geq d(s, v)$.
- **Claim 2:** When a vertex is marked **sure**, $d[v] = d(s, v)$.

- **Claims 1 and 2** imply the **theorem**.



What have we learned?

- Dijkstra's algorithm finds shortest paths in weighted graphs with non-negative edge weights.
- Along the way, it constructs a nice tree.
 - We could post this tree in COB1!
 - Then people would know how to get places quickly.

YOINK!

