

On How to Learn, Do and Write Theory

**Spring 2026
Brown University**

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How Theory is (Often) Taught

How Theory is (Often) Taught

1. Here is **problem X**.
2. Here is **method A**.
3. Therefore **solution**

How to Solve Theory Problems (?)

1. Write down the **problem X**.
2. Think *real* hard.
3. Write down the **solution**.

≈Murray Gell-Mann



How Theory is Done

Simplification



Active



How Theory Problems are Solved

1. Isolate a toy **model case x** of major **problem X**.
2. Solve **model case x** using **method A**.
3. Try using **method A** to solve the full **problem X**.
4. This does not succeed but **method A** can be extended to **model cases x' and x''**.
5. Eventually, it is realized that **method A** relies crucially on a **property P** being true which holds for **model cases x, x' and x''**.
6. Conjecture that **property P** is true for all instances of **problem X**.
7. Discover a family of **counterexamples y, y', y'',...** to this conjecture.
8. Take the simplest **counterexample y** in this family, and try to solve **problem X** for this special case. Meanwhile, try to see whether **method A** can work without **property P**.
9. Discover several counterexamples in which **method A** fails, in which the cause of failure can be definitely traced back to **property P**. Abandon efforts to modify **method A**.
10. Realize that **counterexample y** is related to a **problem Z** in another field.

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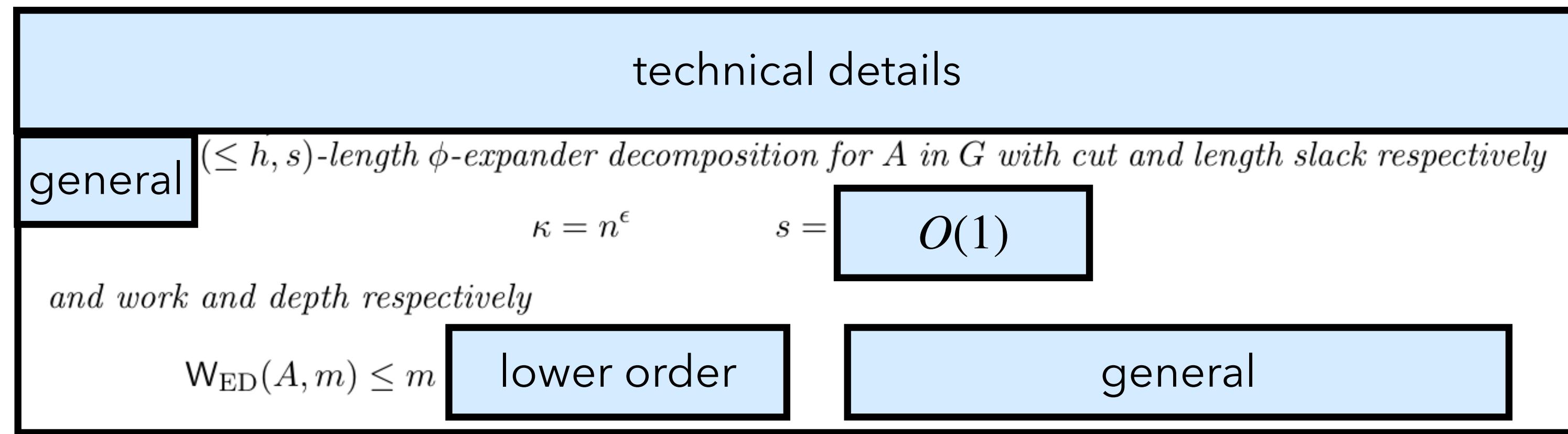
22. **Method Z** is rapidly developed and extended to get the **solution** to **problem X**.

≈Terry Tao

How to Learn Theory

How to Learn Theory

Simplification



ignore lower order parameters / technical details

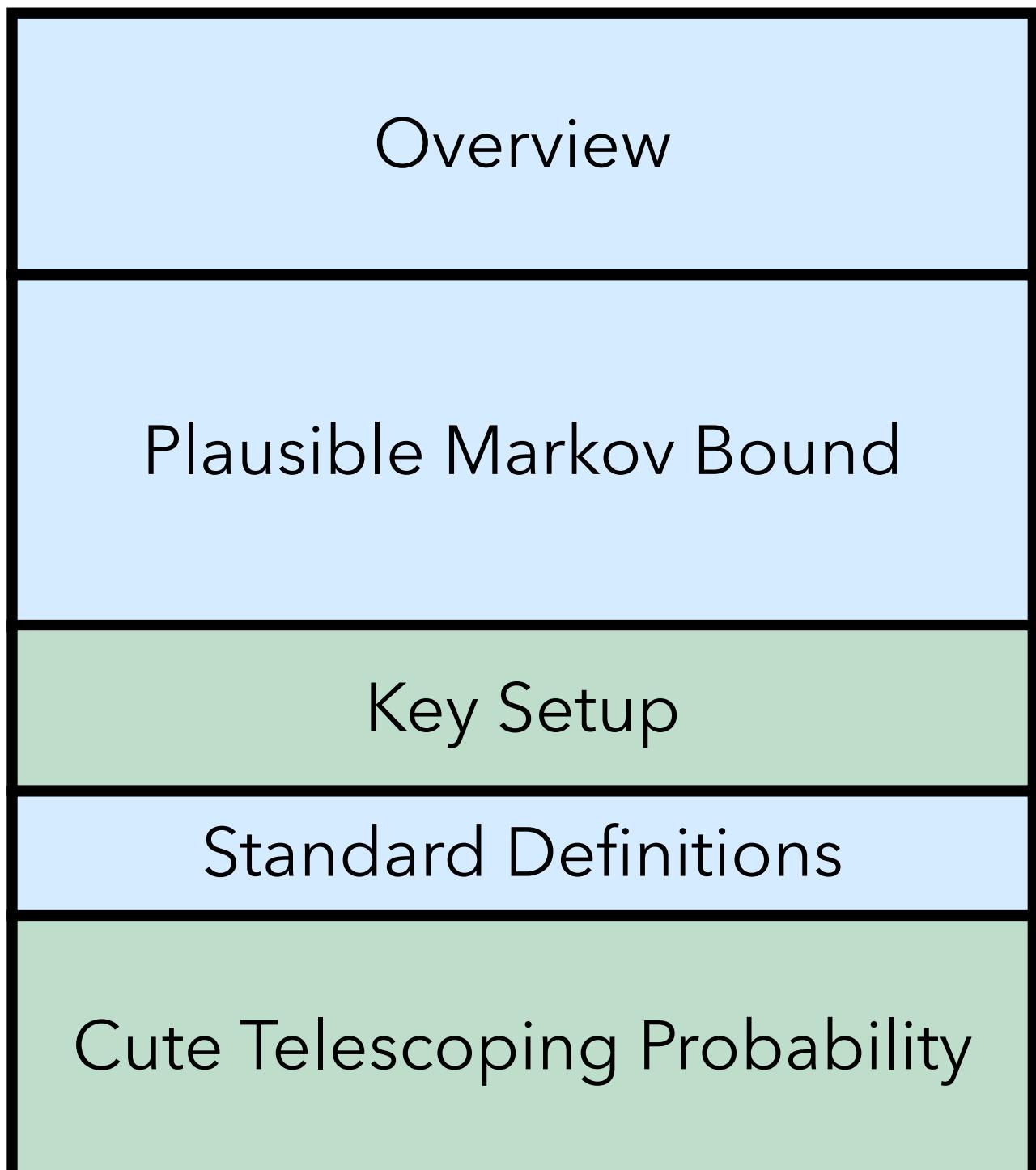
fix parameters

apply theorem to special cases

Simplify theorems

How to Learn Theory

Simplification

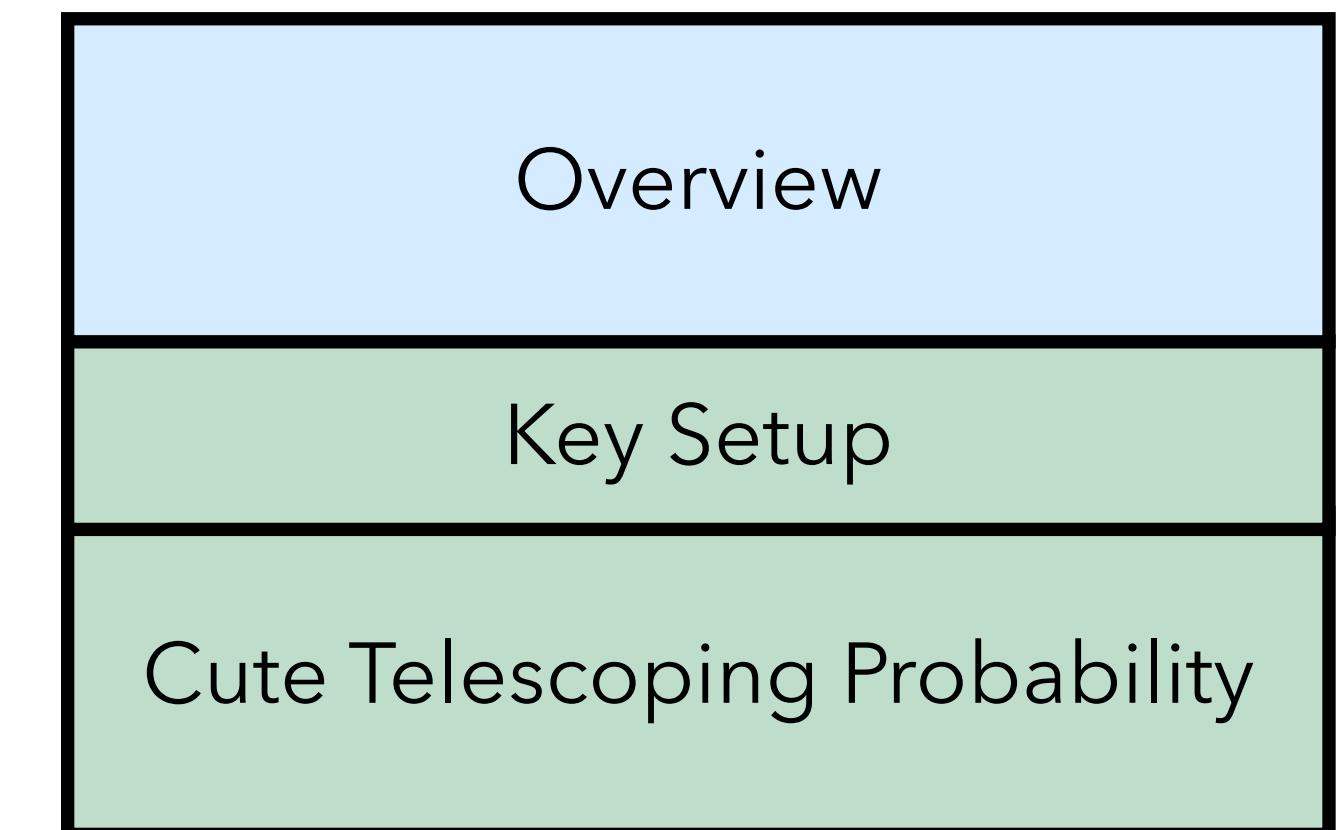


Proof on Arbitrary Graphs

*skip standard +
plausible details*

note tricks

*do proof on
special cases*



Proof on Regular Graphs

Simplify proofs

How to Learn Theory

Active Engagement

Proof of Lemma 15.3.5: We're trying to analyze $\Pr[S_{iw} = 1 | X_{iw} = 1]$ for every $w \in V$. To do this, let's order V by distance to $\{u, v\}$, so

$$d(w_i, \{u, v\}) \leq d(w_{i+1}, \{u, v\})$$

for all i .

Now let's fix some w_j , and suppose that w_j cuts $\{u, v\}$ at level i , i.e., $|B(w_j, r_{i-1}) \cap \{u, v\}| = 1$. Then by the definition of our ordering, every w_k with $k < j$ must have $|B(w_k, r_{i-1}) \cap \{u, v\}| > 0$. Thus if *any* of these nodes come before w_j in π , we know that w_j will not settle u, v at level i , since at least one of u, v will have already been clustered by the time w_j gets to form clusters. Since π is a random permutation, the probability that w_j comes before the w_k for all $k < j$ is exactly $1/j$. Thus $\Pr[S_{iwj} = 1 | X_{iwj} = 1] \leq 1/j$. So by setting $b_{wj} = 1/j$, we have proved the first part of the lemma.

5

The proof of the second part of the lemma is now straightforward:

$$\sum_{w \in V} b_w = \sum_{j=1}^n b_{wj} = \sum_{j=1}^n \frac{1}{j} = H_n = O(\log n),$$

as claimed. ■

Proof of Lemma 15.3.6: Now we're trying to prove that $\sum_{i=0}^{\log \Delta} 2^{i+3} \Pr[X_{iw} = 1] \leq 16d(u, v)$ for all $w \in V$. Without loss of generality, let's assume that $d(w, u) \leq d(w, v)$. In order for w to cut u, v at level i (i.e., for $X_{iw} = 1$), it needs to be the case that $r_{i-1} \in [d(w, u), d(w, v))$. Moreover, r_{i-1} is distributed uniformly in $[2^{i-2}, 2^{i-1}]$. Thus

$$\Pr[X_{iw} = 1] = \frac{|[2^{i-2}, 2^{i-1}] \cap [d(w, u), d(w, v)]|}{|[2^{i-2}, 2^{i-1}]|} = \frac{|[2^{i-2}, 2^{i-1}] \cap [d(w, u), d(w, v)]|}{2^{i-2}}.$$

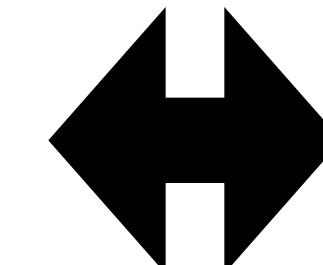
So we have that

$$2^{i+3} \Pr[X_{iw} = 1] = \frac{2^{i+3}}{2^{i-2}} |[2^{i-2}, 2^{i-1}] \cap [d(w, u), d(w, v)]| = 32 |[2^{i-2}, 2^{i-1}] \cap [d(w, u), d(w, v)]|.$$

Thus

$$\sum_{i=0}^{\log \Delta} 2^{i+3} \Pr[X_{iw} = 1] \leq \sum_{i=0}^{\log \Delta} 32 |[2^{i-2}, 2^{i-1}] \cap [d(w, u), d(w, v)]| = 32 |[d(w, u), d(w, v)]| = 32(d(w, v) - d(w, u)) \leq 32d(u, v),$$

where the final inequality is from the triangle inequality. ■



2 Subclaims

(A) $\sum_{i=0}^{\log \Delta} 2^{i+3} \Pr[X_{iw} = 1] \leq 16d_{uv}$

(B) \exists bound for s.t. $\Pr[S_{iw} | X_{iw}] \leq b_w$ and $\sum_w b_w \leq O(\log n)$

Proving (A)

WLOG suppose $d_{uw} \leq d_{vw} \Rightarrow$
 Notice X_{iw} is if we $B(u, r_i)$ but $v \notin B(u, r_i)$
 $\Rightarrow \Pr[X_{iw}] = \Pr(d_{uw} \leq r_i < d_{vw})$
 Since r_i chosen uniformly in $[2^{i-1}, 2^i]$ this is as $\Pr[X_{iw}] = \Pr(d_{uw} \leq 2^{i-1} < d_{vw}) = \Pr(d_{uw} \leq 2^{i-1})$
 Thus $\sum_i 2^{i+3} \Pr[X_{iw}] = \sum_i 2^{i+3} \Pr[X_{iw}] \approx \sum_i \frac{\log d_{uv}}{2^i} = \Theta\left(\sum_{i=0}^{\log d_{uv}} 2^i - \frac{\log d_{uv}}{2^{\log d_{uv}}}\right)$
 $= \Theta(2(d_{uv} - d_{uv}))$
 $= 16d_{uv}$

Proving (B)

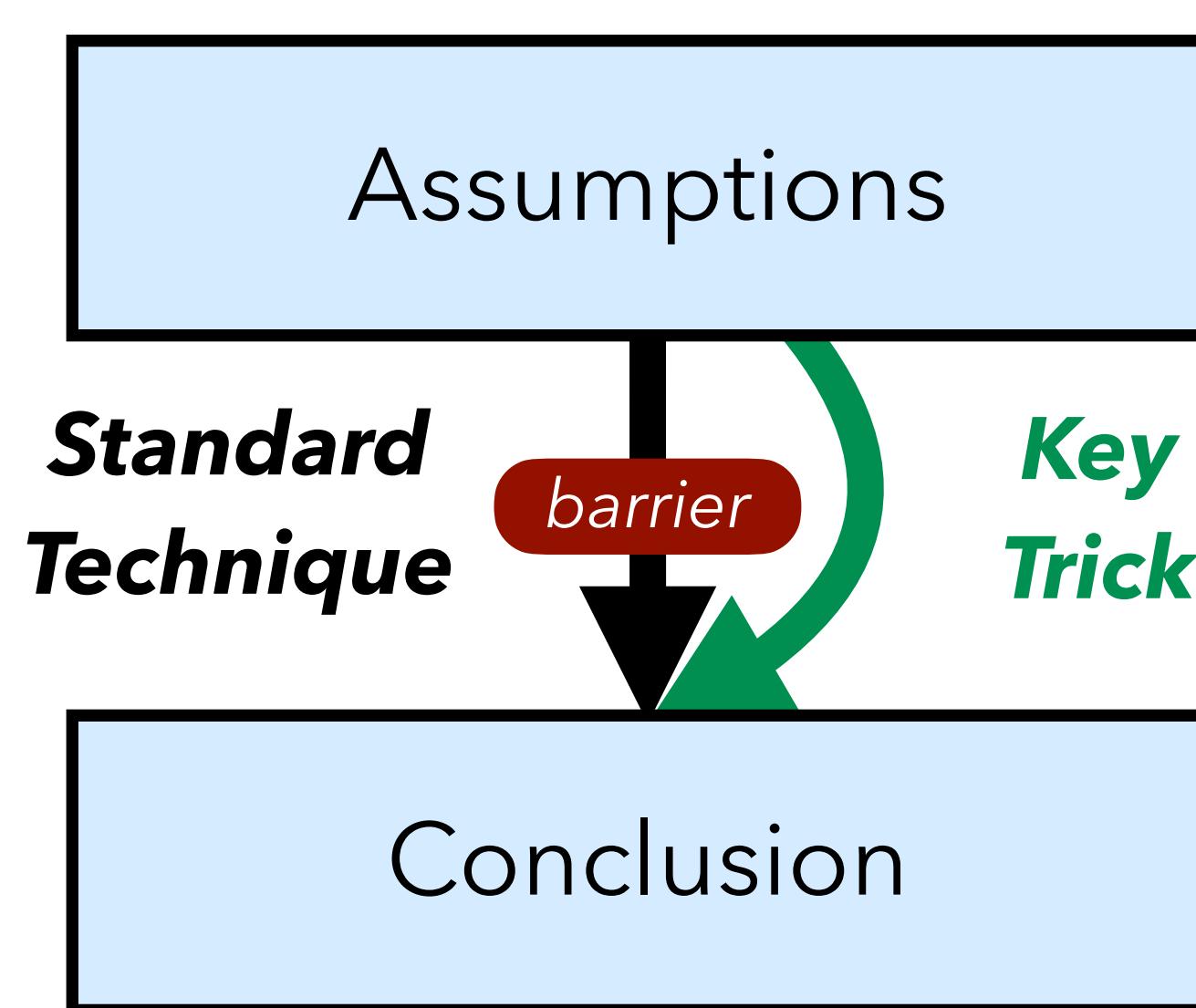
Order w by closeness to (u, v) (i.e. $\min(d_{uw}, d_{vw})$)

Conditioning on a vertex cutting (u, v) . A vertex w settles (u, v) only if $\pi(w) < \pi(u) \wedge w$ closer to (u, v) than u .
 Thus if w is the 5th closest vertex and we consider projecting π onto the 5 closest vertices the w not always come first \Rightarrow happens w/ Pr $\frac{1}{5}$.
 So let $b_w = \frac{1}{j}$ and so $\Pr[S_{iw} | X_{iw}] \leq \frac{1}{j}$.
 Sum of a harmonic gives $\sum_w b_w \leq O(\log n)$

Recreate Proofs after you learn them; see where you get stuck

How to Learn Theory

Active Engagement



Invent Stories that you like / will remember

Do the Same for LC Expander Decompositions?

While G has an (h, s) -length ϕ -sparse cut C :

- (h, s) -length $\tilde{O}(\phi)$ -sparse
- β -balanced

50

Problem 1: Union of Sparse LC Not Clearly Sparse

Definition: (h, s) -length cut C is ϕ -sparse if there is an h -length unit demand D of size $|C|/\phi$ that it hs -separates

Sum of witnessing demands is not unit!

57

Union of Sparse LC Cuts is Sparse

Goal: transform witness demands into a separated unit demand

Insight: demand graph is an s -parallel greedy graph

Theorem[HHT]: s -parallel greedy graphs have arboricity at most $\tilde{O}(n^{1/s})$

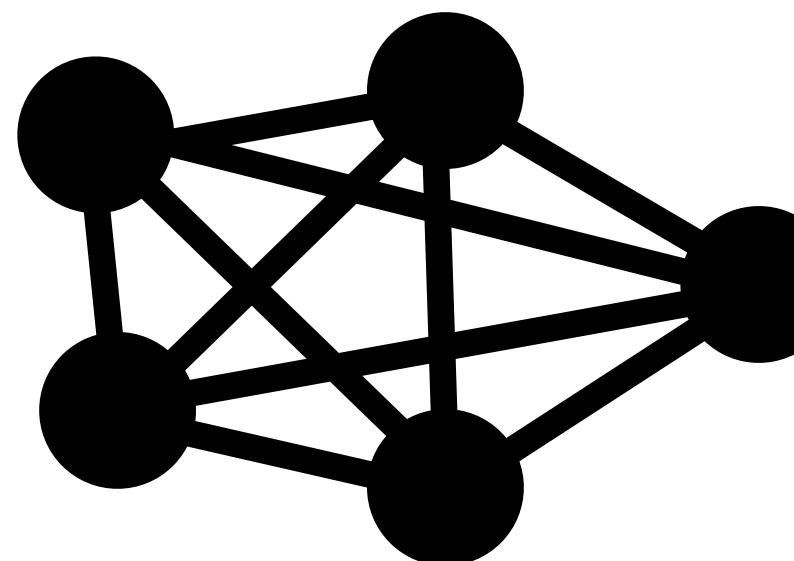
Original (Non-Unit) Demand

Dispersed (Unit) Demand

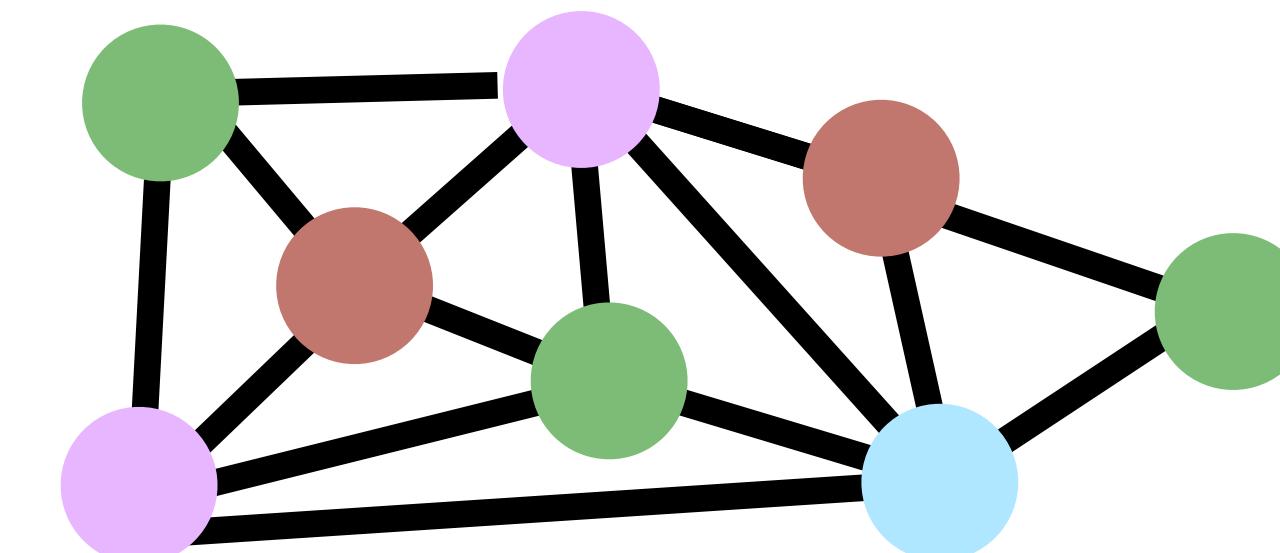
How to Learn Theory

Active Engagement

Theorem: Every planar graph is 4-colorable

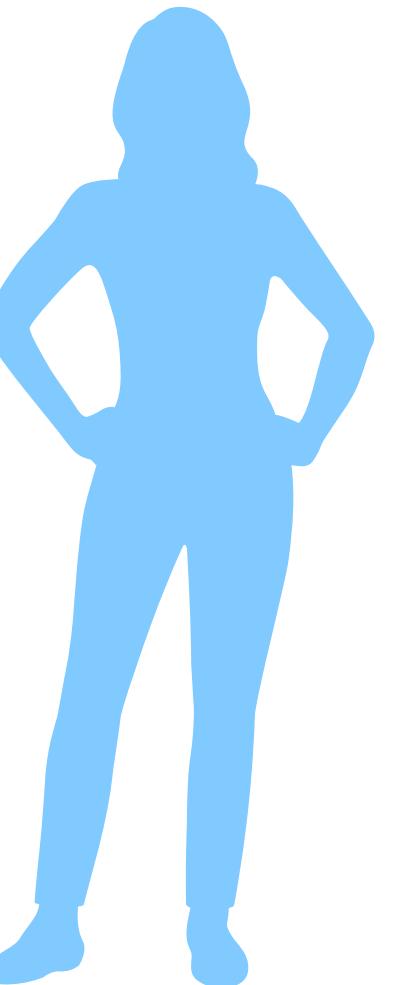


why assumptions needed?



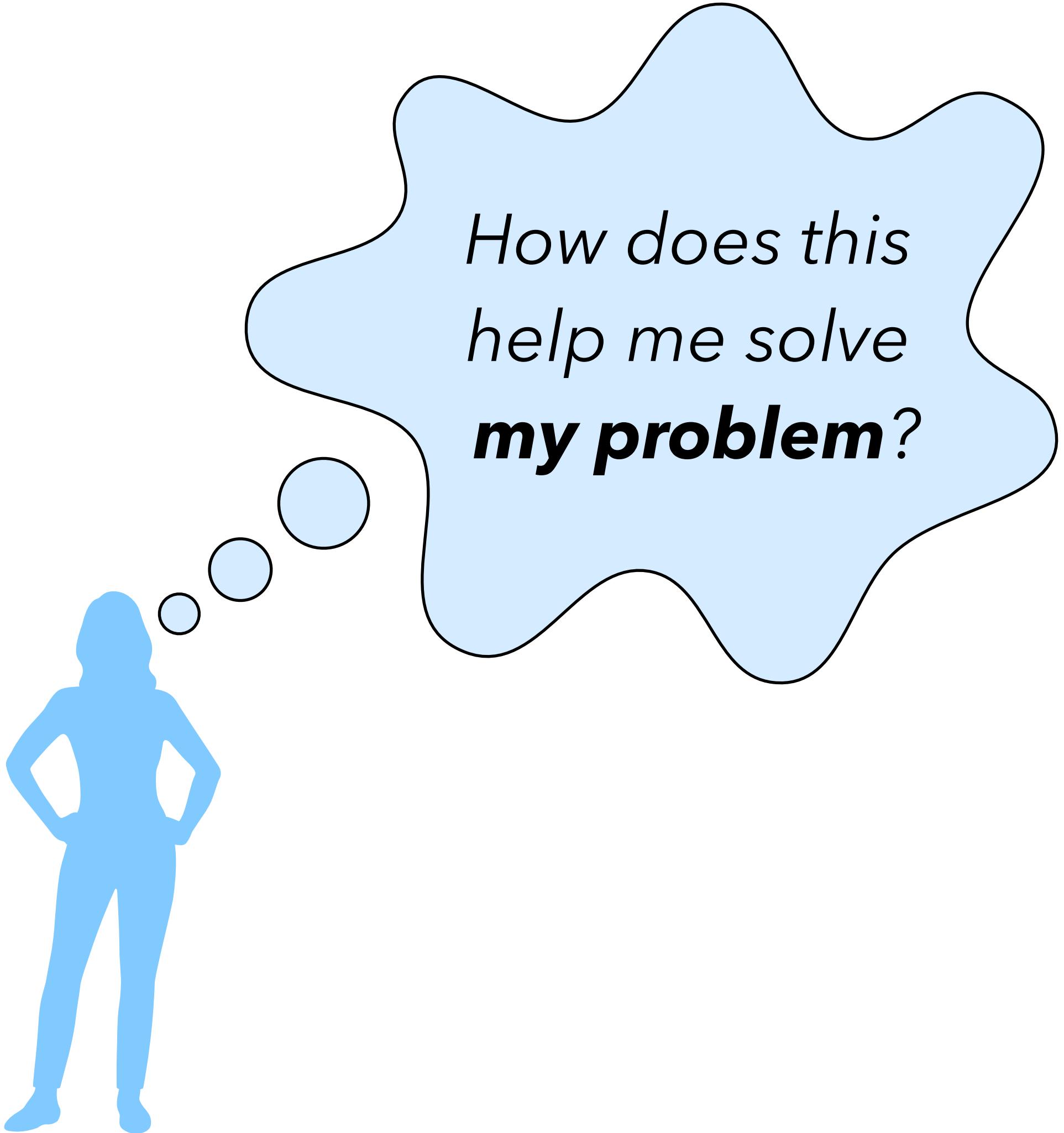
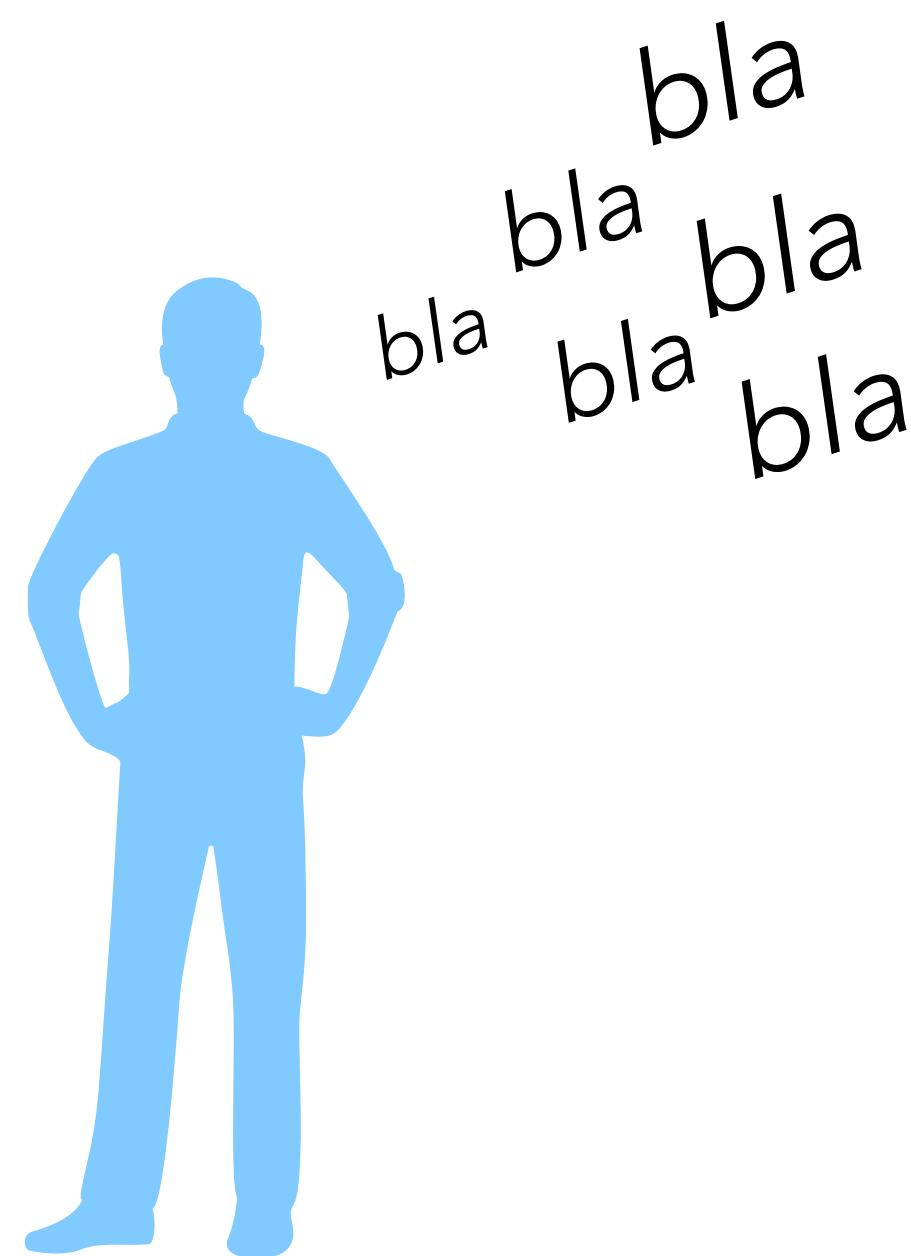
what does this give on e.g.s?

Ask Yourself Questions



How to Learn Theory

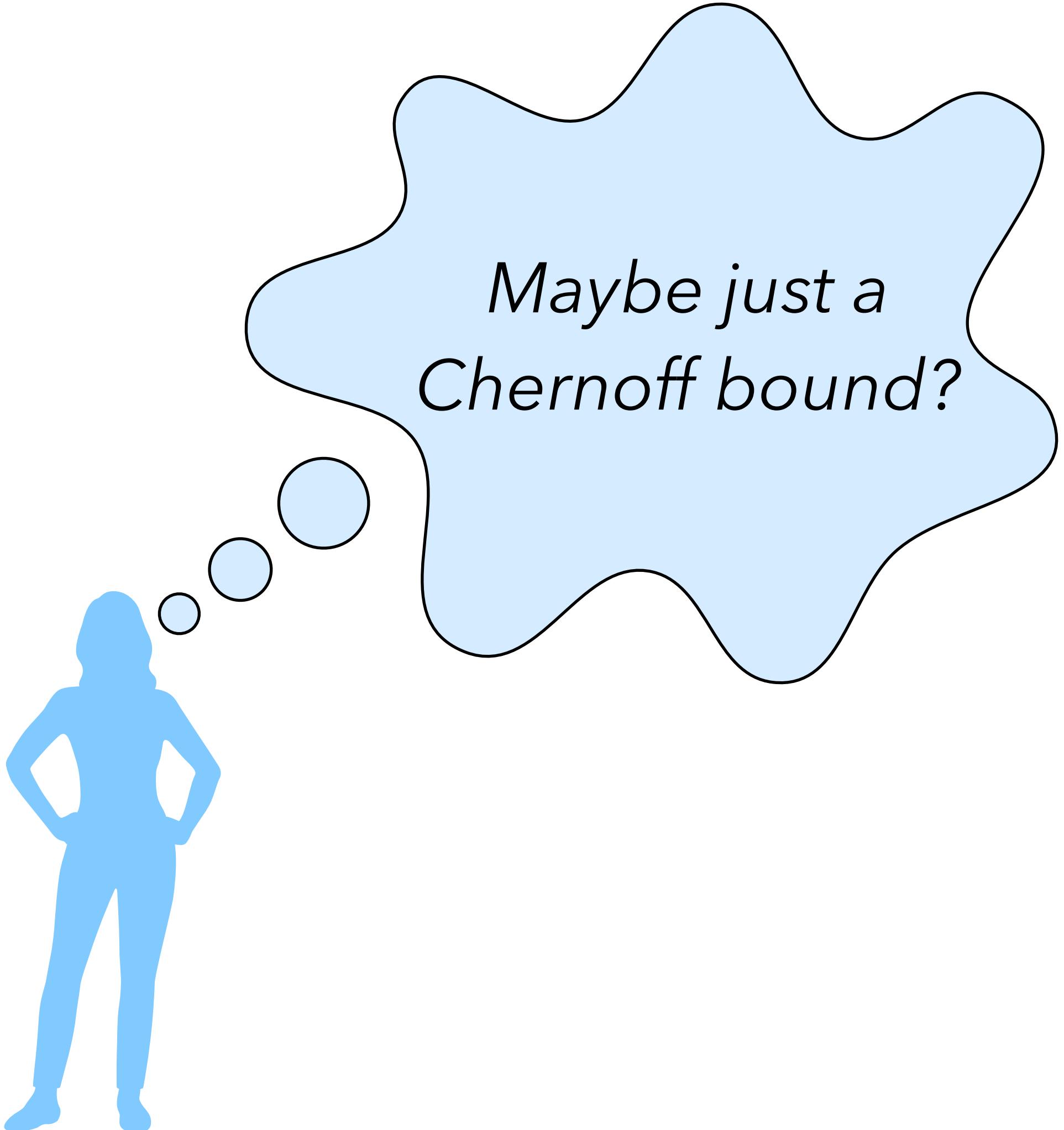
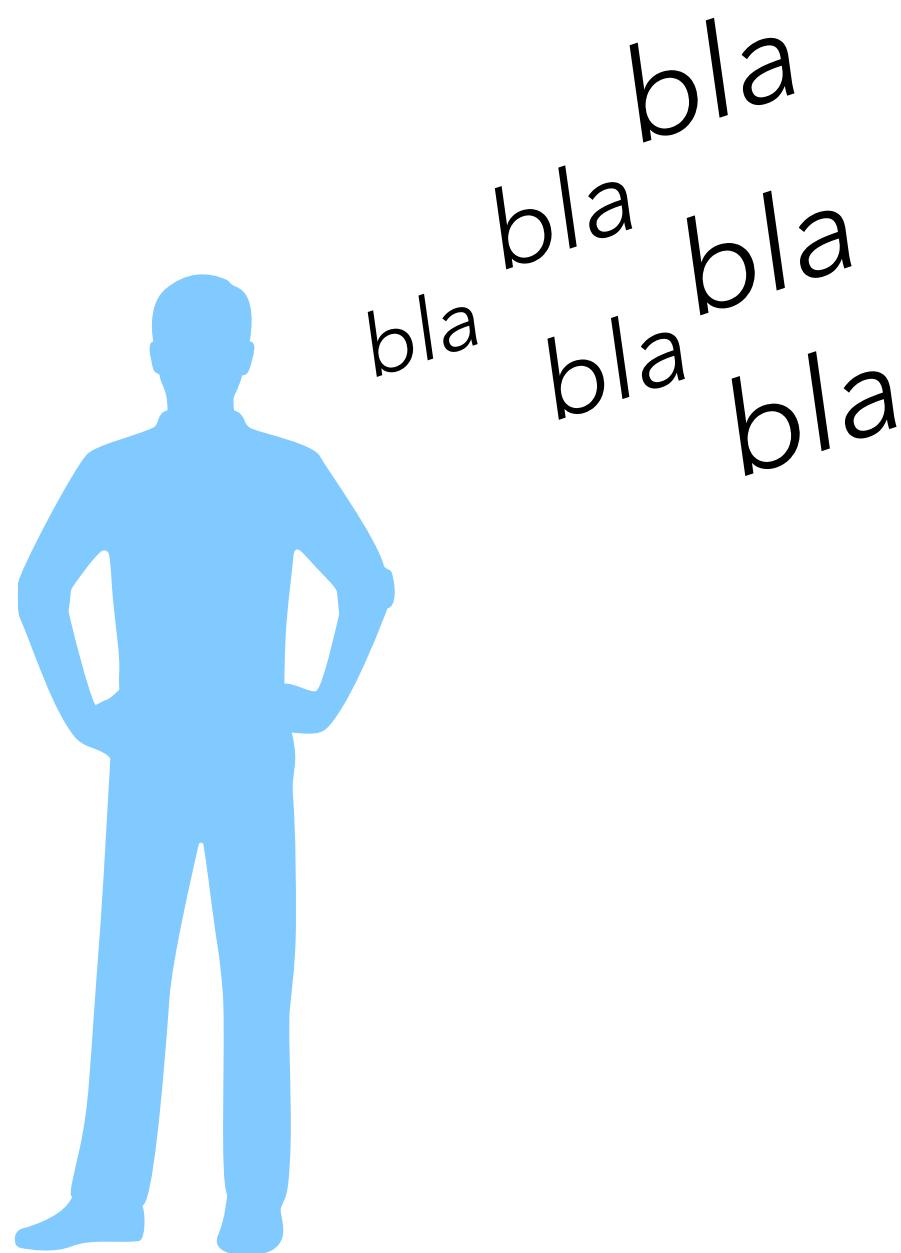
Active Engagement



Anchor Your Learning with a problem you like

How to Learn Theory

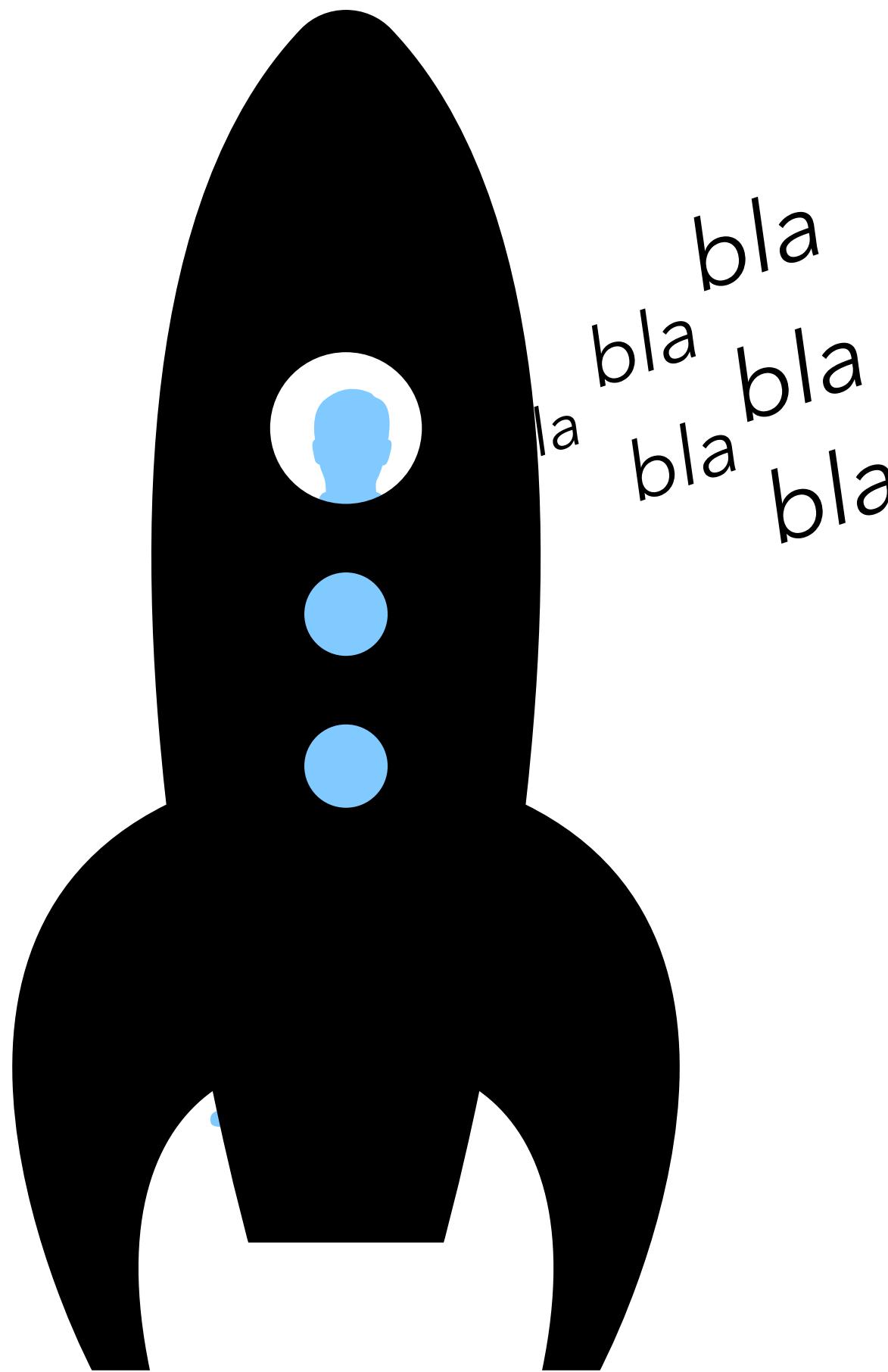
Active Engagement



Guess what's coming next

How to Learn Theory

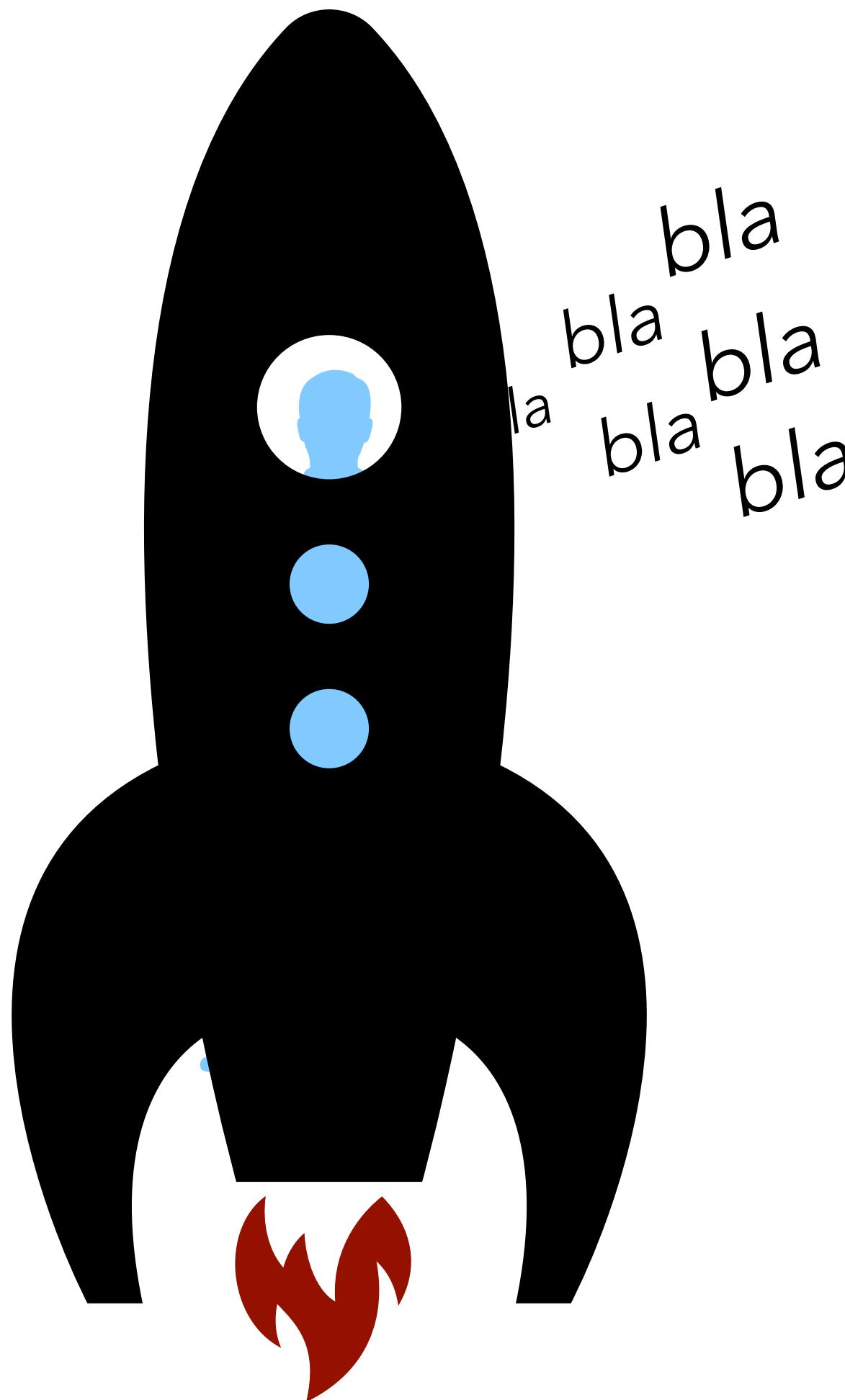
Active Engagement



Ask Questions if you're confused

How to Learn Theory

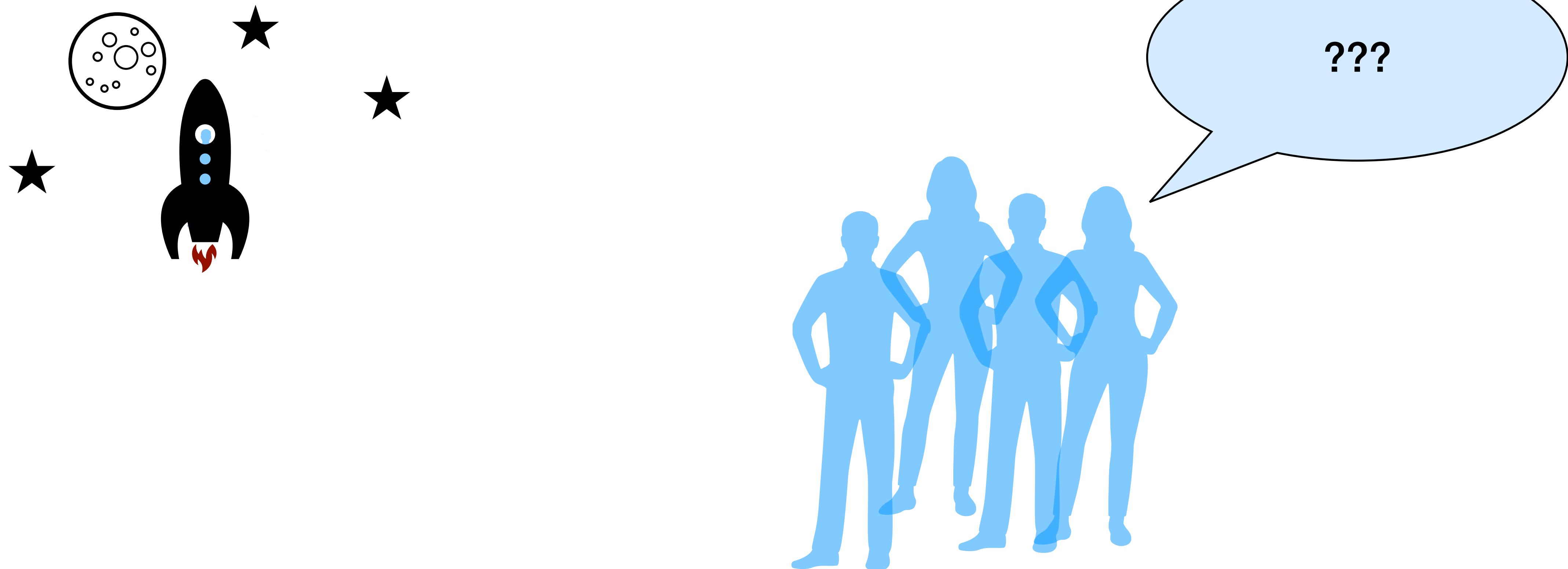
Active Engagement



Ask Questions if you're confused

How to Learn Theory

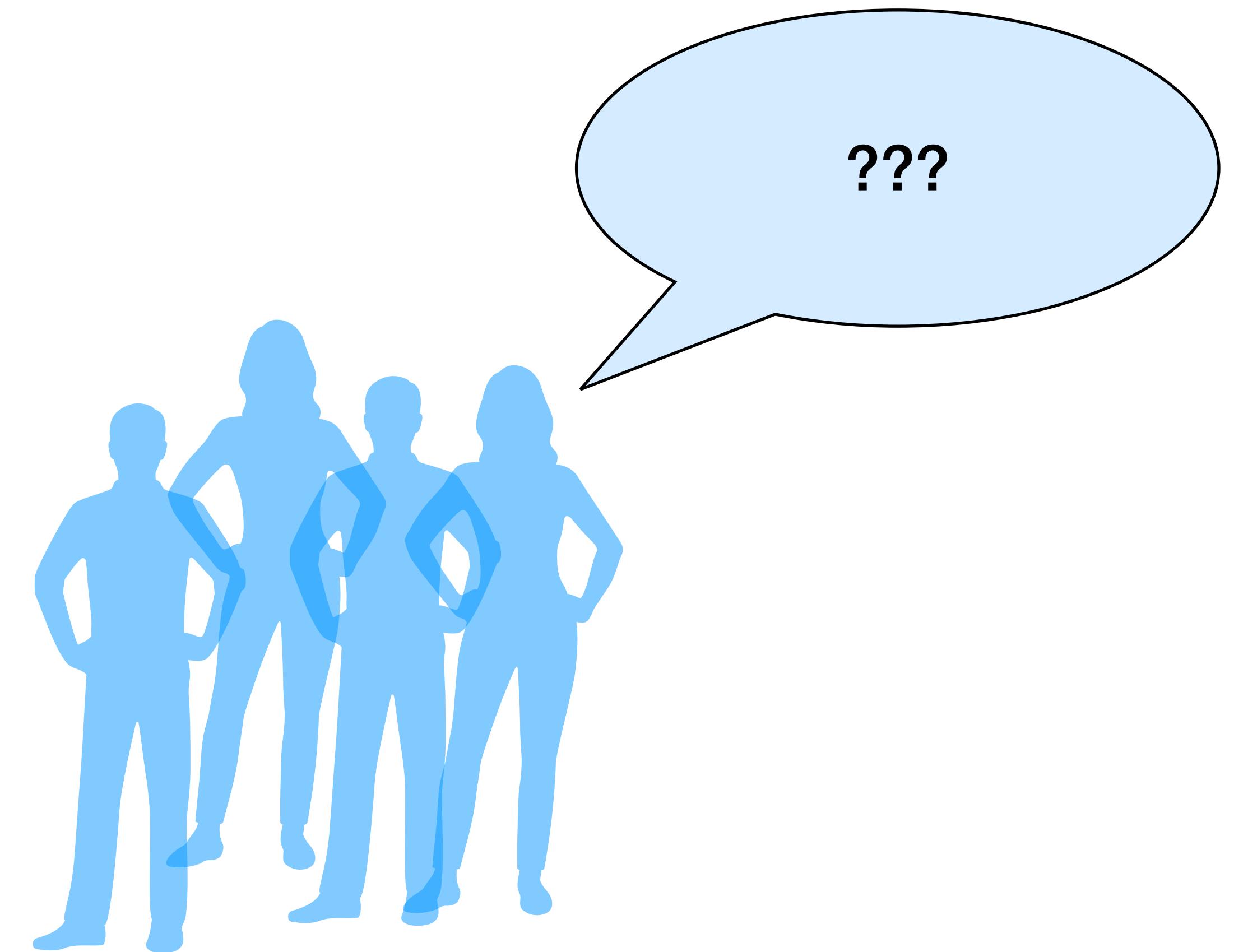
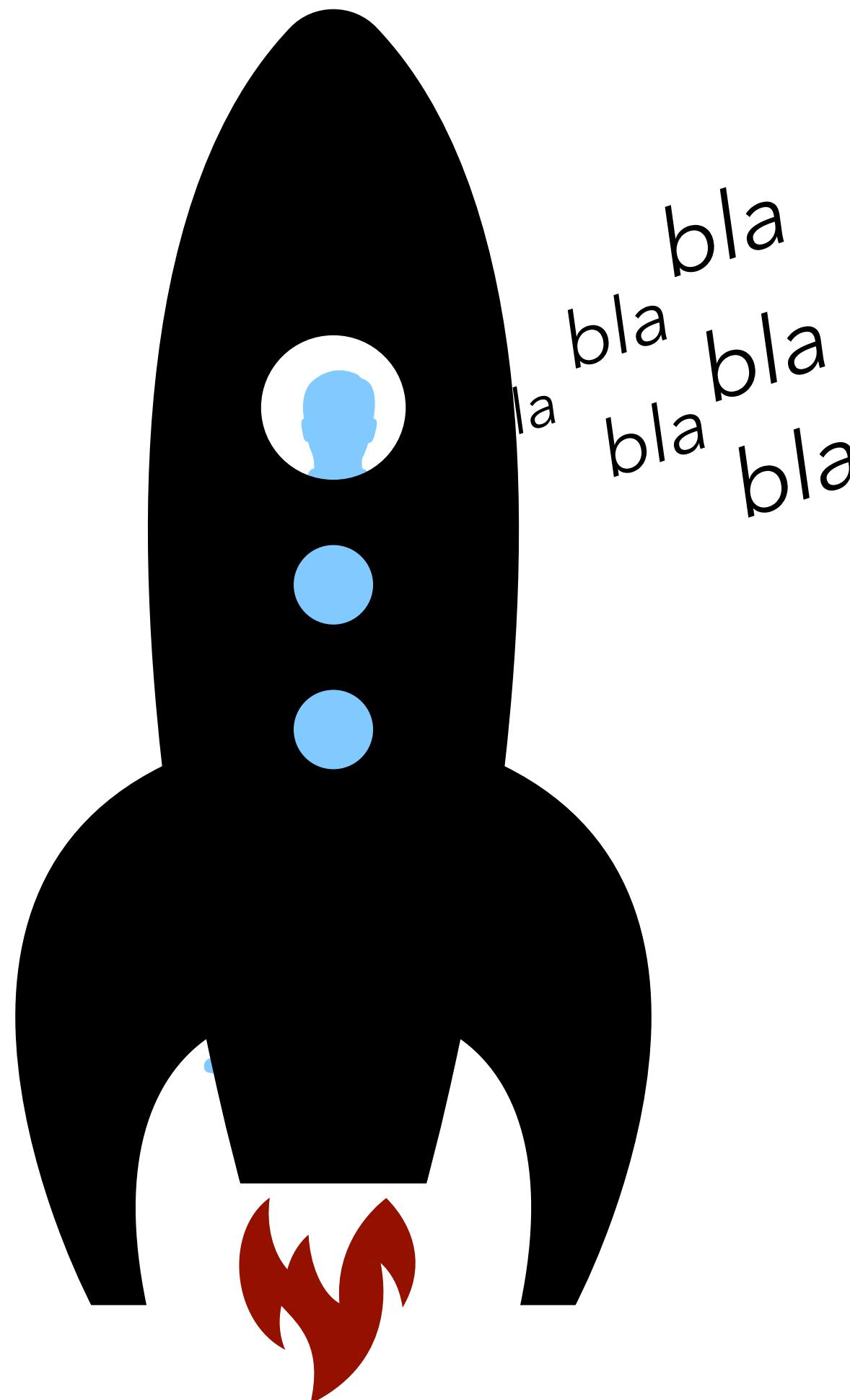
Active Engagement



Ask Questions if you're confused

How to Learn Theory

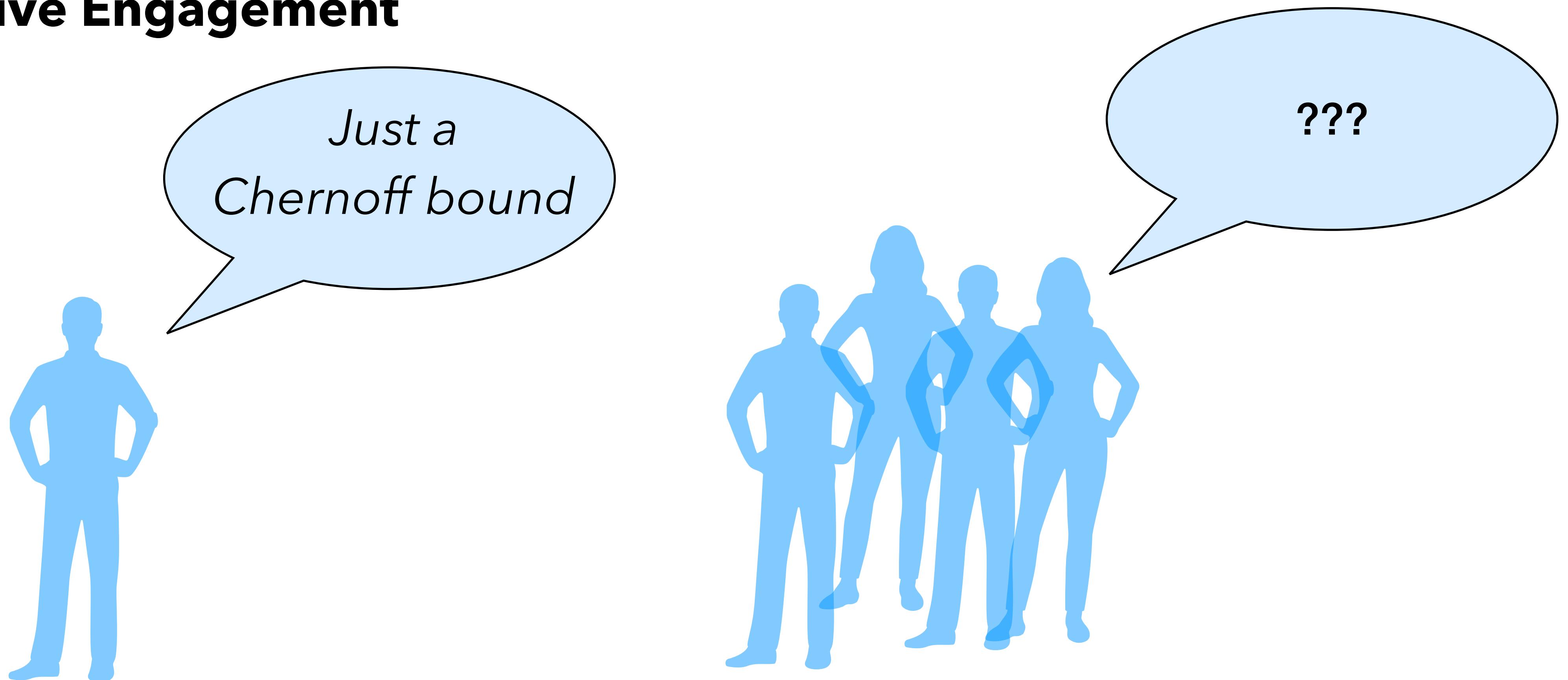
Active Engagement



Ask Questions if you're confused

How to Learn Theory

Active Engagement



Ask Questions if you're confused

How to Learn Theory

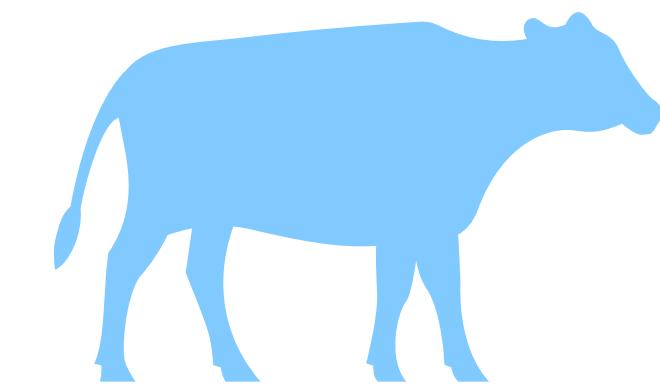
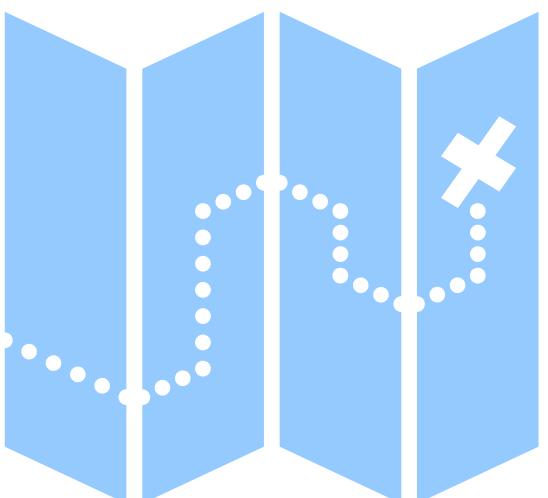
What to Aim For

Roadmaps of proof

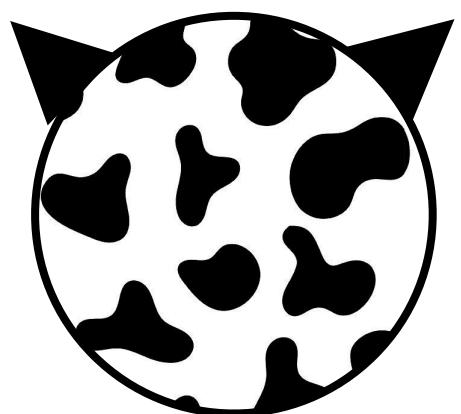
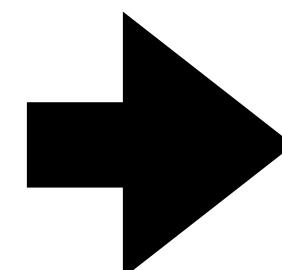


Tools and stories you'll remember

Intuition of how to think about complexity



a cow



a cow
(up to constants)

How to Do Theory

How to Do Theory

Doing Theory is Hard

Can Succeed with a Wide Range of Aptitudes:

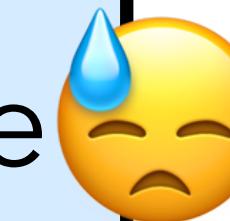
- Mathematical **quick**ness 
- **Good memory** 
- Good ~~intuition~~ 
- Reliable 
- Just really **curious** 
- **Stick-To-Itiveness** 
- **Impatient** / only interested in elegant solutions 
- ...

Work to Your Strengths

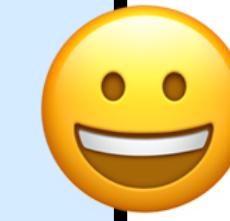
How to Do Theory

You Will Get Stuck

Theorem: Every planar graph is 4-colorable



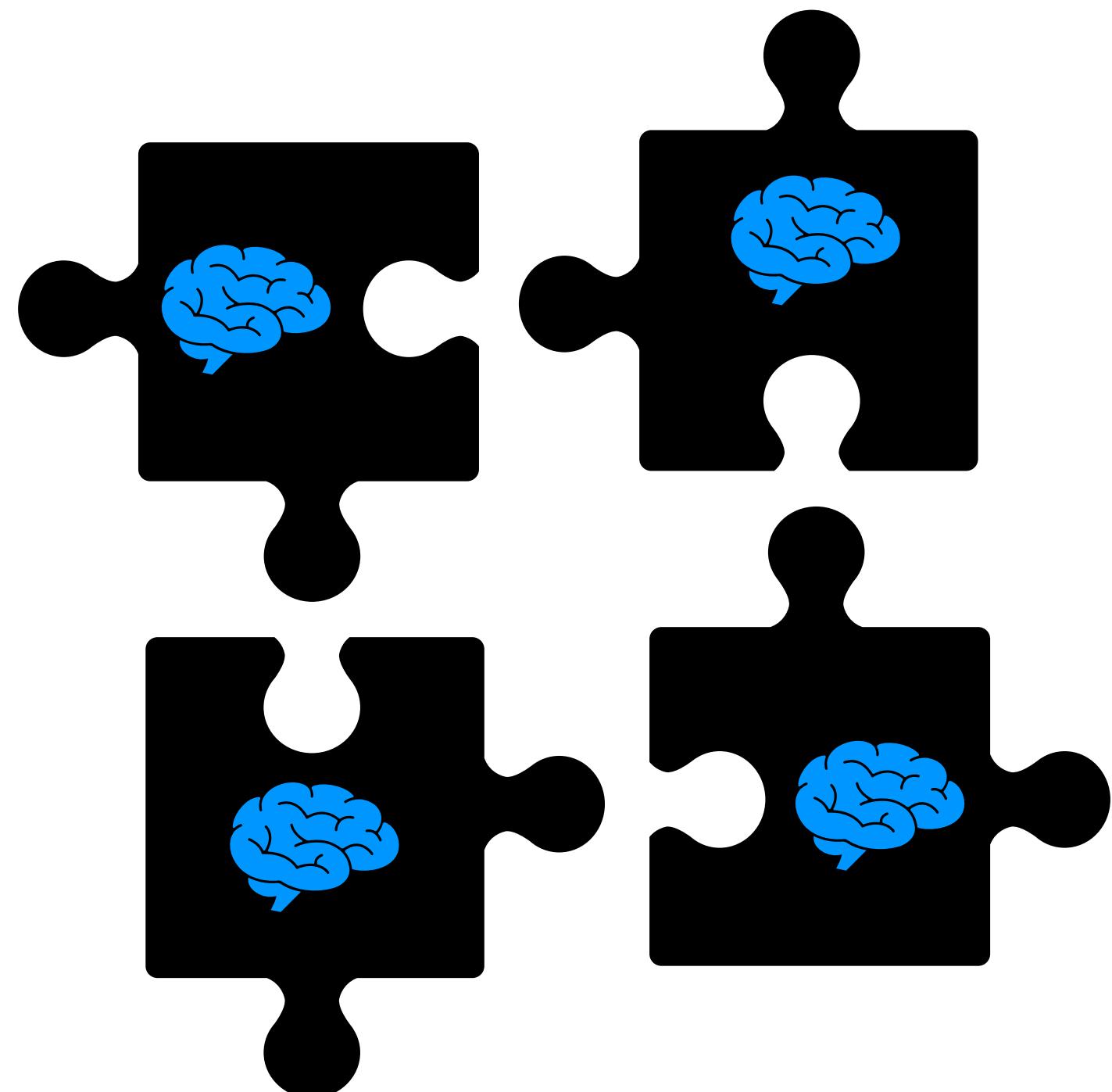
Theorem: Every tree is 4-colorable



Simplify your problem

How to Do Theory

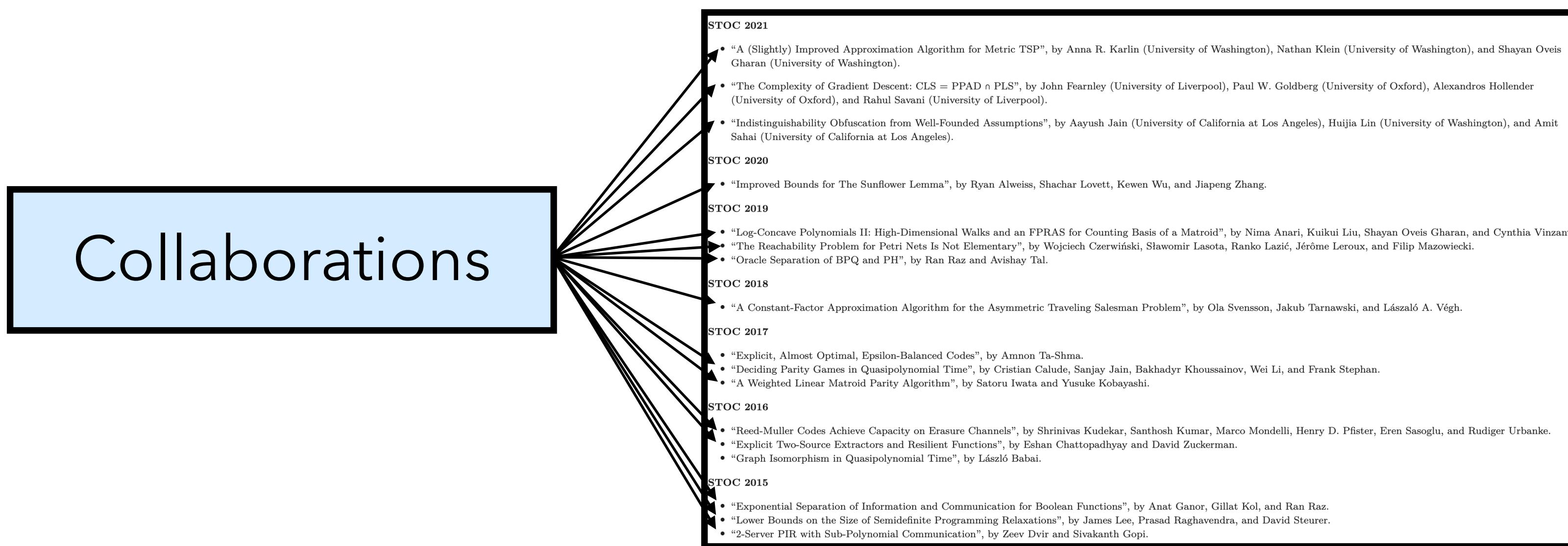
You Will Get Stuck



Collaborate

How to Do Theory

You Will Get Stuck

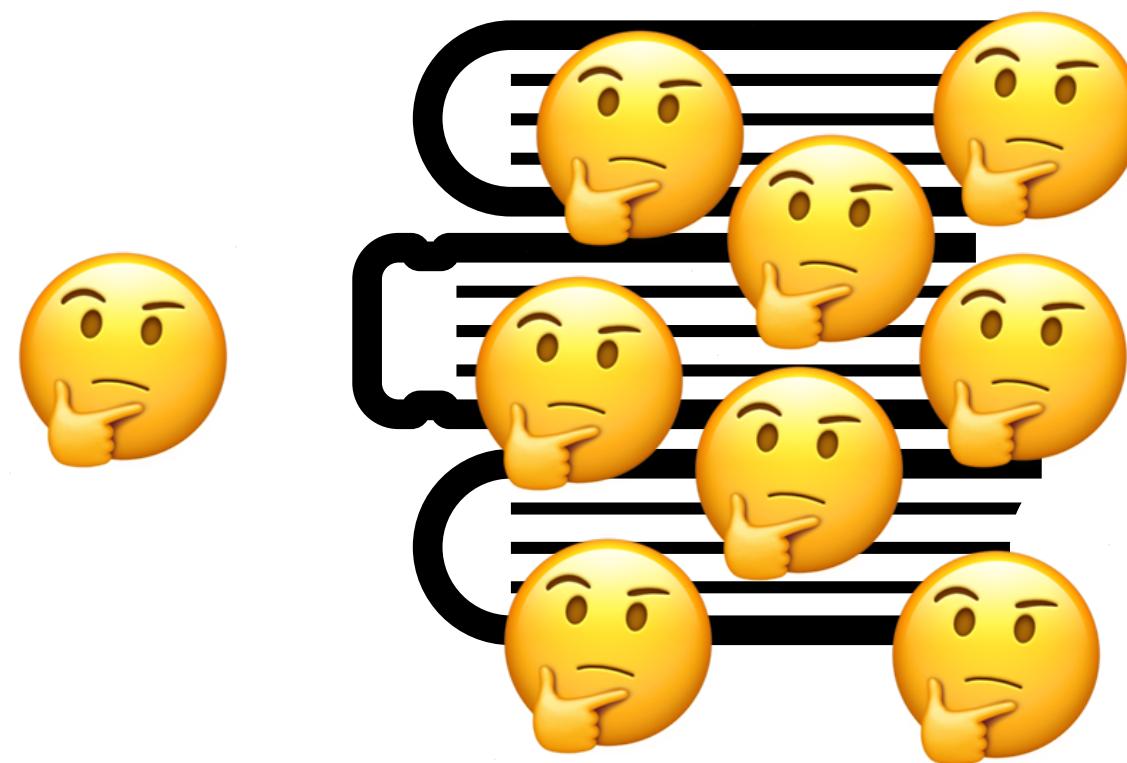


STOC Best Papers

Collaborate

How to Do Theory

You Will Get Stuck



Read Related Work

How to Do Theory

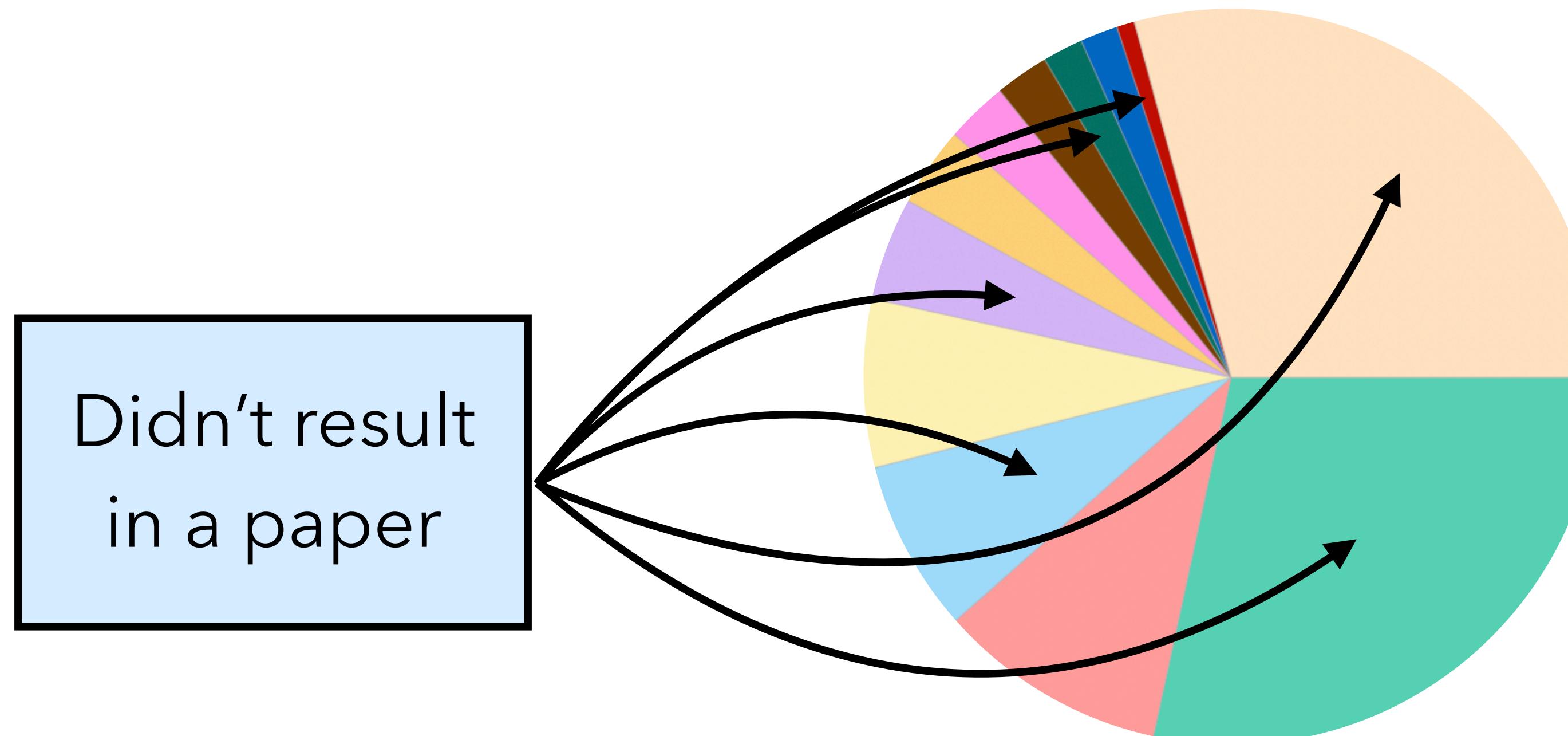
You Will Get Stuck



Cut Yourself Slack

How to Do Theory

A Few Mantras

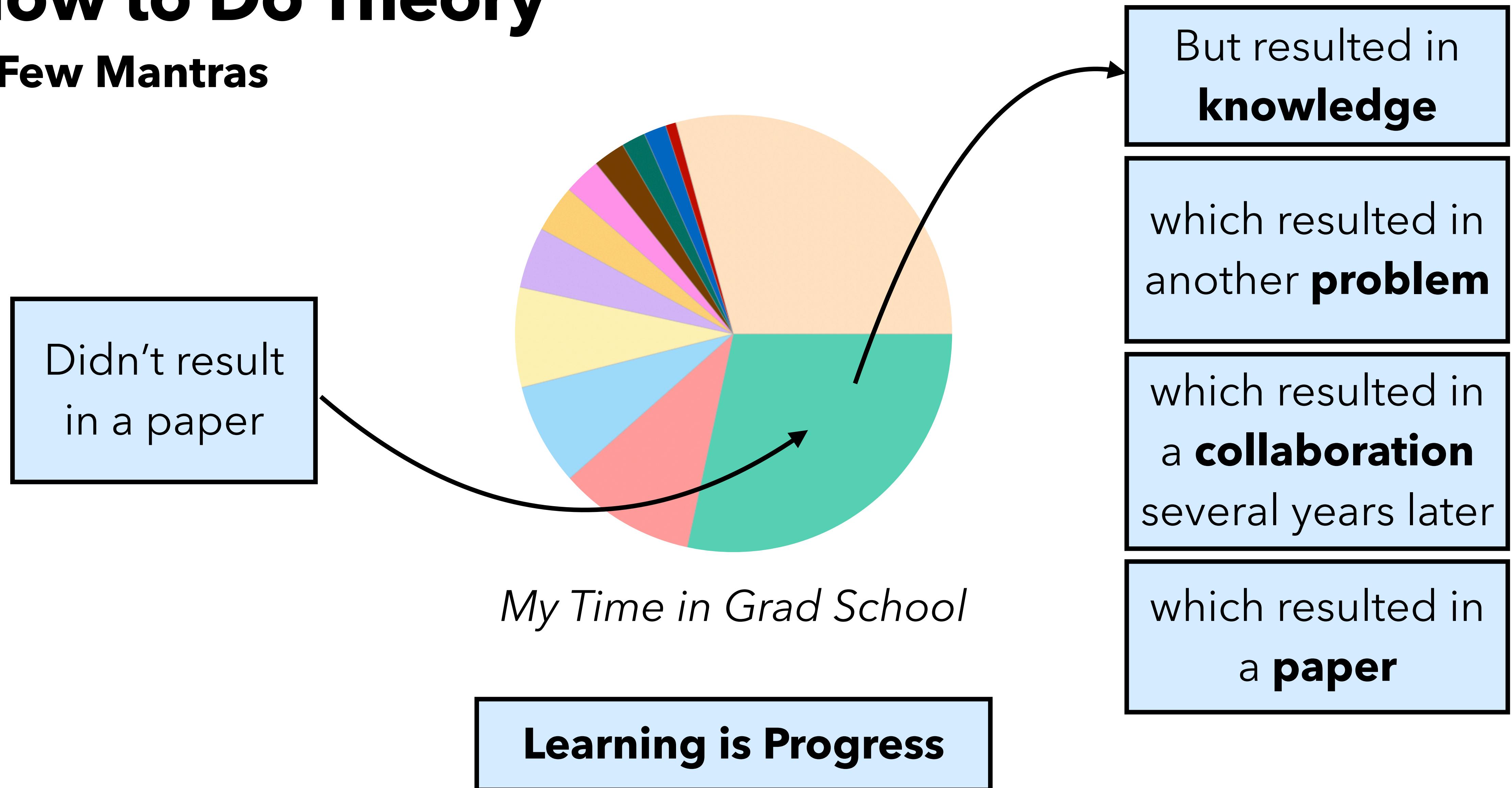


My Time in Grad School

Failure is Common

How to Do Theory

A Few Mantras



How to Do Theory

A Few Mantras

How Theory Problems are Solved

1. Isolate a toy **model case x** of major **problem X**.
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9. Discover several counterexamples in which **method A** fails, in which the cause of failure can be definitely traced back to **property P**. Abandon efforts to modify **method A**.
10. Realize that **counterexample y** is related to a **problem Z** in another field.

• • •

22. **Method Z** is rapidly developed and extended to get the **solution** to **problem X**.

Any New Insight is Progress

How to Do Theory

A Few Mantras

Grades



Often

Awards



↓

Internships



→



Undergrad

Theorems
You
Prove



Infrequent

Paper
You
Write



Grad Student

**How Cool
Theory Is**



Often

Learn to Love the Process Not the Outcome

How to Write Theory

Writing Dos and Don'ts

Bad References



Theorem 1: $1+1+1=3$

Proof:

First we show $1+1=2$...

Next, we show $2+1=3$...

Theorem 2: $2+1+1=4$

Proof:

First we show $1+1=2$...

Next, we show $2+2=4$...



Lemma: $1+1=2$

Theorem 1: $1+1+1=3$

Proof:

By Lemma $1+1=2$

Next, we show $2+1=3$...

Theorem 2: $2+1+1=4$

Proof:

By Lemma $1+1=2$

Next, we show $2+2=4$...

Abstract out reused arguments into lemmas

Writing Dos and Don'ts

Bad References



Theorem 1: $1+1+1=3$

Proof:

First we show $1+1=2$...

Next, we show $2+1=3$...

Theorem 2: $2+1+1=4$

Proof:

By the argument in Theorem 1, $1+1=2$

Next, we show $2+2=4$...



Lemma: $1+1=2$

Theorem 1: $1+1+1=3$

Proof:

By Lemma $1+1=2$...

Next, we show $2+1=3$...

Theorem 2: $2+1+1=4$

Proof:

By Lemma $1+1=2$...

Next, we show $2+2=4$...

Don't reference the insides of other proofs

Writing Dos and Don'ts

Bad References



Theorem 1: $1+1+1=3$

Proof:

Hershkowitz et al. showed that $1+1=2\dots$

Next, we show $2+1=3\dots$



Lemma[Hershkowitz et al.]: $1+1=2$

Theorem 1: $1+1+1=3$

Proof:

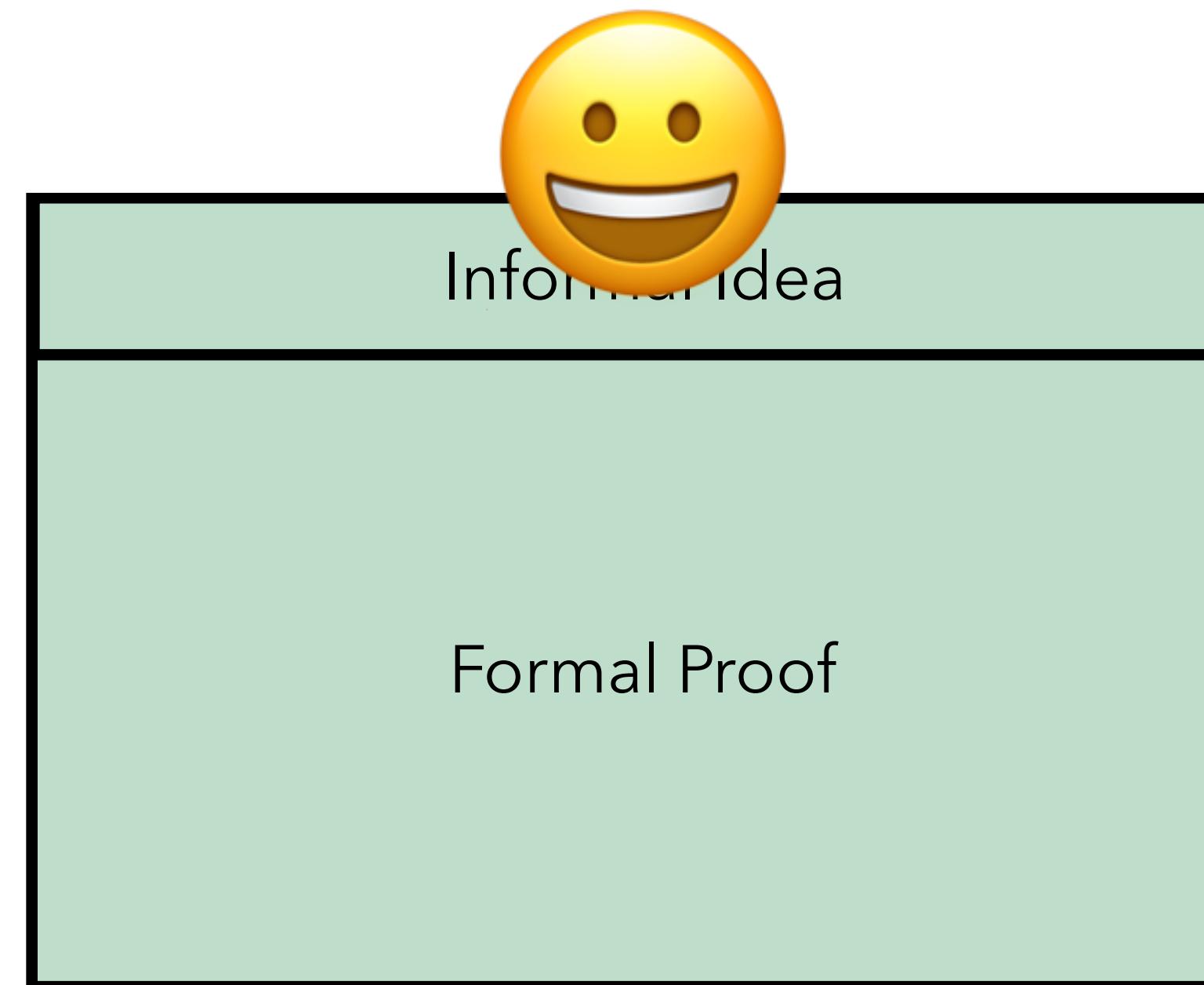
By Lemma $1+1=2\dots$

Next, we show $2+1=3\dots$

Don't reference facts not stated as theorems/lemmas/etc.

Writing Dos and Don'ts

Intuition



Give intuition / an overview at the beginning of your proofs

Writing Dos and Don'ts

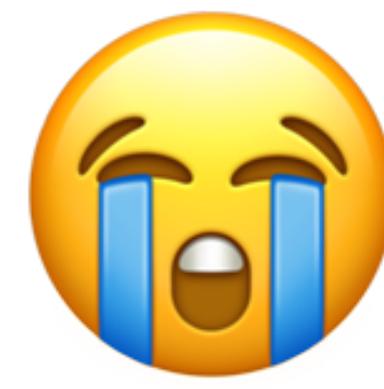
Intuition



Hand Waving



Just Right

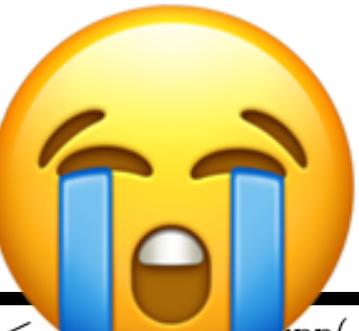


Excruciating
Formality

Balance intuition and formality

Writing Dos and Don'ts

General Style



and so $\sum_{S,i} \sum_k |\text{supp}(A_{Si,k} \cup B_{Si,k})| \leq \sum_{S,i} |\text{supp}(A_S^{(i)})| \leq \tilde{O}(n^{1+O(\epsilon)} + N^{O(\epsilon)} L^2)$. Thus, plugging this bound on $\sum_{S,i} |\text{supp}(A_{Si,k} \cup B_{Si,k})|$ into the guarantees of [Theorem 10.4](#) and the fact that our pairs are $L \cdot N^{O(\epsilon)}$ -batchable, we have that each time we compute a cutmatch in [step 2](#), the total number of edges we add across all G_S for $S \in \mathcal{N}[h']$ for a fixed h' is at most $\tilde{O}(m + L \cdot N^{O(\epsilon)} + n^{1+O(\epsilon)} + N^{O(\epsilon)} L) = \tilde{O}(m + n^{1+O(\epsilon)} + L^2 \cdot N^{O(\epsilon)})$. Since we have $1/\epsilon$ iterations, it follows that the number of edges across all G_S for $S \in \mathcal{N}[h']$ is never more than $\frac{1}{\epsilon} \cdot \tilde{O}(m + n^{1+O(\epsilon)} + L^2 \cdot N^{O(\epsilon)})$. It follows that the work and depth to compute all cut strategies for all $S \in \mathcal{N}[h']$ for all $1/\epsilon$ -many iterations and all $h' \leq h \cdot \frac{1}{\epsilon} \cdot (s)^{O(1/\epsilon)}$ a power of 2 in [step 2](#) are respectively $\frac{1}{\epsilon} \cdot \sum_i W_{\text{cut-strat}}(A_i, m_i)$ and $\frac{1}{\epsilon} \cdot \max_i D_{\text{cut-strat}}(A_i, m_i)$ where $|A_i| \leq |A|/L$ for all i and $\sum_i m_i \leq \tilde{O}(m + n^{1+O(\epsilon)} + L^2 \cdot N^{O(\epsilon)})$.



and so

$$\sum_{S,i} \sum_k |\text{supp}(A_{Si,k} \cup B_{Si,k})| \leq \sum_{S,i} N^{O(\epsilon)} |\text{supp}(A_S^{(i)})| \leq \tilde{O}(n^{1+O(\epsilon)} + N^{O(\epsilon)} L^2)$$

Thus, plugging this bound on $\sum_{S,i} |\text{supp}(A_{Si,k} \cup B_{Si,k})|$ into the guarantees of [Theorem 10.4](#) and the fact that our pairs are $L \cdot N^{O(\epsilon)}$ -batchable, we have that each time we compute a cutmatch in [step 2](#), the total number of edges we add across all G_S for $S \in \mathcal{N}[h']$ for a fixed h' is at most

$$\tilde{O}(m + L \cdot N^{O(\epsilon)} + n^{1+O(\epsilon)} + N^{O(\epsilon)} L) = \tilde{O}(m + n^{1+O(\epsilon)} + L^2 \cdot N^{O(\epsilon)}).$$

Since we have $1/\epsilon$ iterations, it follows that the number of edges across all G_S for $S \in \mathcal{N}[h']$ is never more than

$$\frac{1}{\epsilon} \cdot \tilde{O}(m + n^{1+O(\epsilon)} + L^2 \cdot N^{O(\epsilon)}).$$

It follows that the work and depth to compute all cut strategies for all $S \in \mathcal{N}[h']$ for all $1/\epsilon$ -many iterations and all $h' \leq h \cdot \frac{1}{\epsilon} \cdot (s)^{O(1/\epsilon)}$ a power of 2 in [step 2](#) are respectively

$$\frac{1}{\epsilon} \cdot \sum_i W_{\text{cut-strat}}(A_i, m_i) \tag{19}$$

and

$$\frac{1}{\epsilon} \cdot \max_i D_{\text{cut-strat}}(A_i, m_i) \tag{20}$$

where $|A_i| \leq |A|/L$ for all i and $\sum_i m_i \leq \tilde{O}(m + n^{1+O(\epsilon)} + L^2 \cdot N^{O(\epsilon)})$.

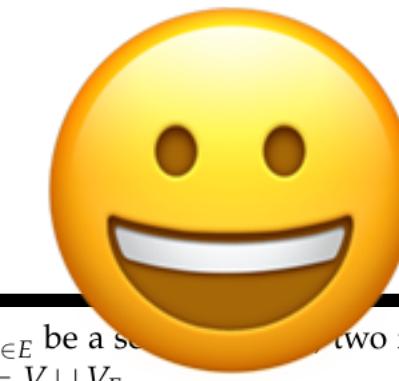
Use whitespace (align*^s) generously

Writing Dos and Don'ts

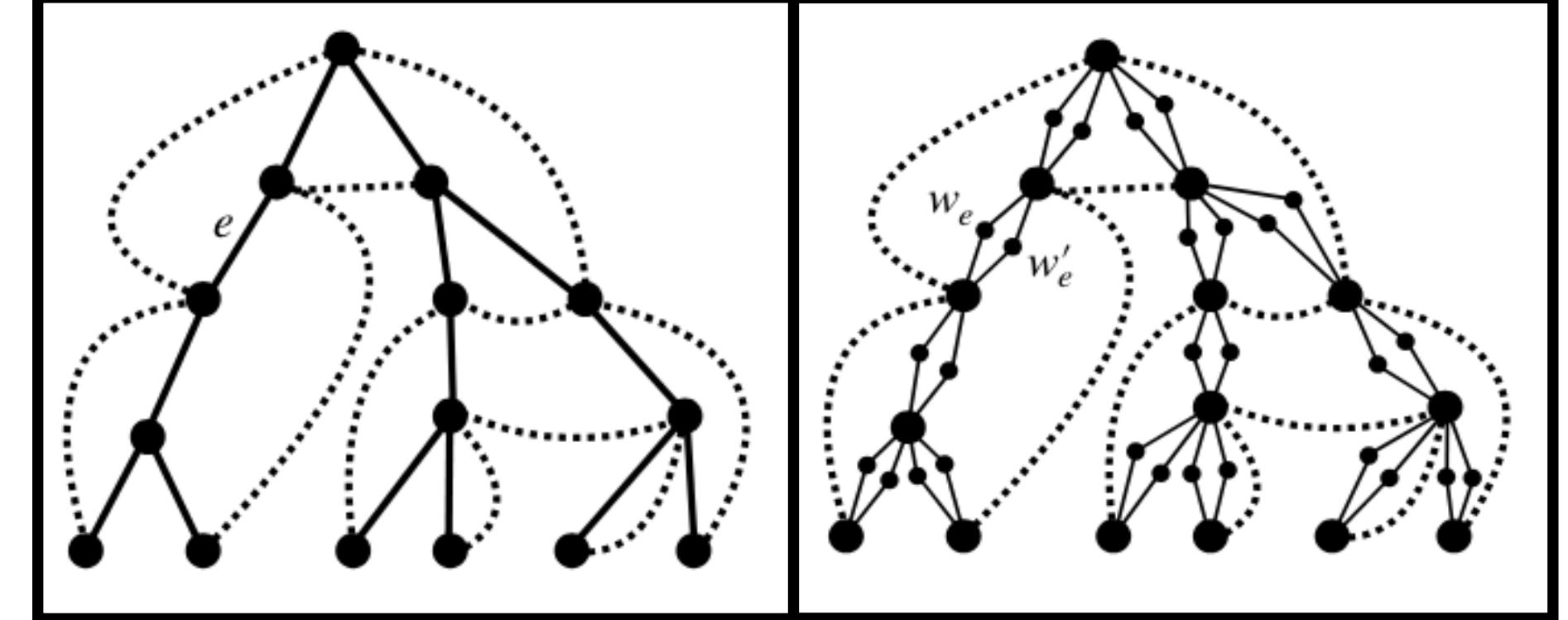
General Style



1. **Vertices:** Let $V_E := \{w_e, w'_e\}_{e \in E}$ be a set of vertices, two for each edge of E . The vertex set of our k -ECSM instance is $W := V \cup V_E$.
2. **Edges:** For each edge $e = \{u, v\} \in E$, we have 4 edges in our k -ECSM instance, namely $\{u, w_e\}$, $\{w_e, v\}$, $\{u, w'_e\}$, and $\{w'_e, v\}$. Let E_{Gadget} be all such edges. The edge set of our k -ECSM instance is $B := E_{\text{Gadget}} \cup L$.
3. **Costs:** The cost of each edge $b \in B$ in our k -ECSM instance is 1, i.e., $c_b = 1$.



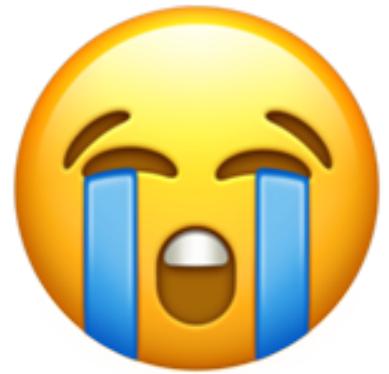
1. **Vertices:** Let $V_E := \{w_e, w'_e\}_{e \in E}$ be a set of vertices, two for each edge of E . The vertex set of our k -ECSM instance is $W := V \cup V_E$.
2. **Edges:** For each edge $e = \{u, v\} \in E$, we have 4 edges in our k -ECSM instance, namely $\{u, w_e\}$, $\{w_e, v\}$, $\{u, w'_e\}$, and $\{w'_e, v\}$. Let E_{Gadget} be all such edges. The edge set of our k -ECSM instance is $B := E_{\text{Gadget}} \cup L$.
3. **Costs:** The cost of each edge $b \in B$ in our k -ECSM instance is 1, i.e., $c_b = 1$.



Use (a lot of) figures

Writing Dos and Don'ts

LaTeX Nits (many courtesy of Ryan O'Donnell)



Quoth the Raven “Nevermore”.



Quoth the Raven ``Nevermore”.

Quoth the Raven ”Nevermore”.

Quoth the Raven “Nevermore”.

Writing Dos and Don'ts

LaTeX Nits (many courtesy of Ryan O'Donnell)



Inner product $\$<\!\!x,y\!\!>\$$.

Inner product $\langle x, y \rangle$.



Inner product $\$\\langle\!\! x,y \\rangle\!\! \$$.

Inner product $\langle x, y \rangle$.

Writing Dos and Don'ts

LaTeX Nits (many courtesy of Ryan O'Donnell)



```
$(\frac{x^2}{y}) \leq z$
```

$$\left(\frac{x^2}{y}\right) \leq z$$



```
\$ \left(\frac{x^2}{y}\right) \leq z $
```

$$\left(\frac{x^2}{y}\right) \leq z$$

Writing Dos and Don'ts

LaTeX Nits (many courtesy of Ryan O'Donnell)



```
$ALG(x) = \log n$,
```

ALG(x) = logn



```
$\text{ALG}(x) = \log n$
```

$\text{ALG}(x) = \log n$

Writing Dos and Don'ts

LaTeX Nits (many courtesy of Ryan O'Donnell)



Let G be a k -connected graph.

Let G be a k -connected graph.



Let $\$G\$$ be a $\$k\$$ -connected graph.

Let G be a k -connected graph.

Writing Dos and Don'ts

LaTeX Nits (many courtesy of Ryan O'Donnell)

```
\begin{align}\label{eq}
A &\leq B \\
&\leq D
\end{align}
so $A \leq C$ by \ref{eq}.
```



```
\begin{align}\label{eq}
A &\leq B \bnonumber \\
&\leq D
\end{align}
so $A \leq C$ by \ref{eq}.
```



We have

$$\begin{aligned} A &\leq B \\ &\leq C \end{aligned} \quad \begin{array}{l} (1) \\ (2) \end{array}$$

so $A \leq C$ by Equation 1.

We have

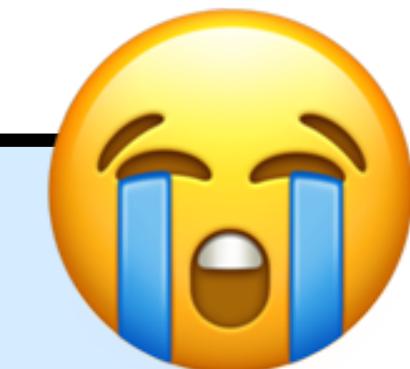
$$\begin{aligned} A &\leq B \\ &\leq C \end{aligned} \quad (1)$$

so $A \leq C$ by Equation 1.

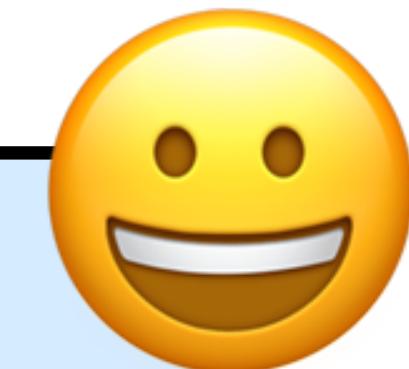
Writing Dos and Don'ts

LaTeX Nits (many courtesy of Ryan O'Donnell)

```
\begin{proof}
\begin{align}
A &\leq B \\
&\leq D
\end{align}
\end{proof}
```



```
\begin{proof} We have
\begin{align}
A &\leq B \\
&\leq D \text{ \b{qedhere}}
\end{align}
\end{proof}
```



Proof.

$$\begin{aligned} A &\leq B \\ &\leq D. \end{aligned}$$

□

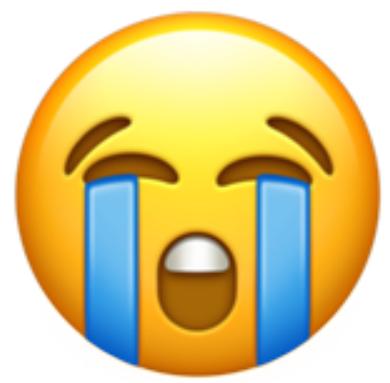
Proof. We have

$$\begin{aligned} A &\leq B \\ &\leq D. \end{aligned}$$

□

Writing Dos and Don'ts

LaTeX Nits (many courtesy of Ryan O'Donnell)



Math is fun, e.g. algebra.

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Math is fun, e.g. \ algebra.

Math is fun, e.g. algebra.

Writing Dos and Don'ts

LaTeX Nits (many courtesy of Ryan O'Donnell)



Math is fun, e.g. algebra.



Math is fun, e.g. \ algebra.

Math is fun, e.g. algebra.
Math is fun, e.g. algebra.



Summary

Learning

Doing

Writing

Simplification



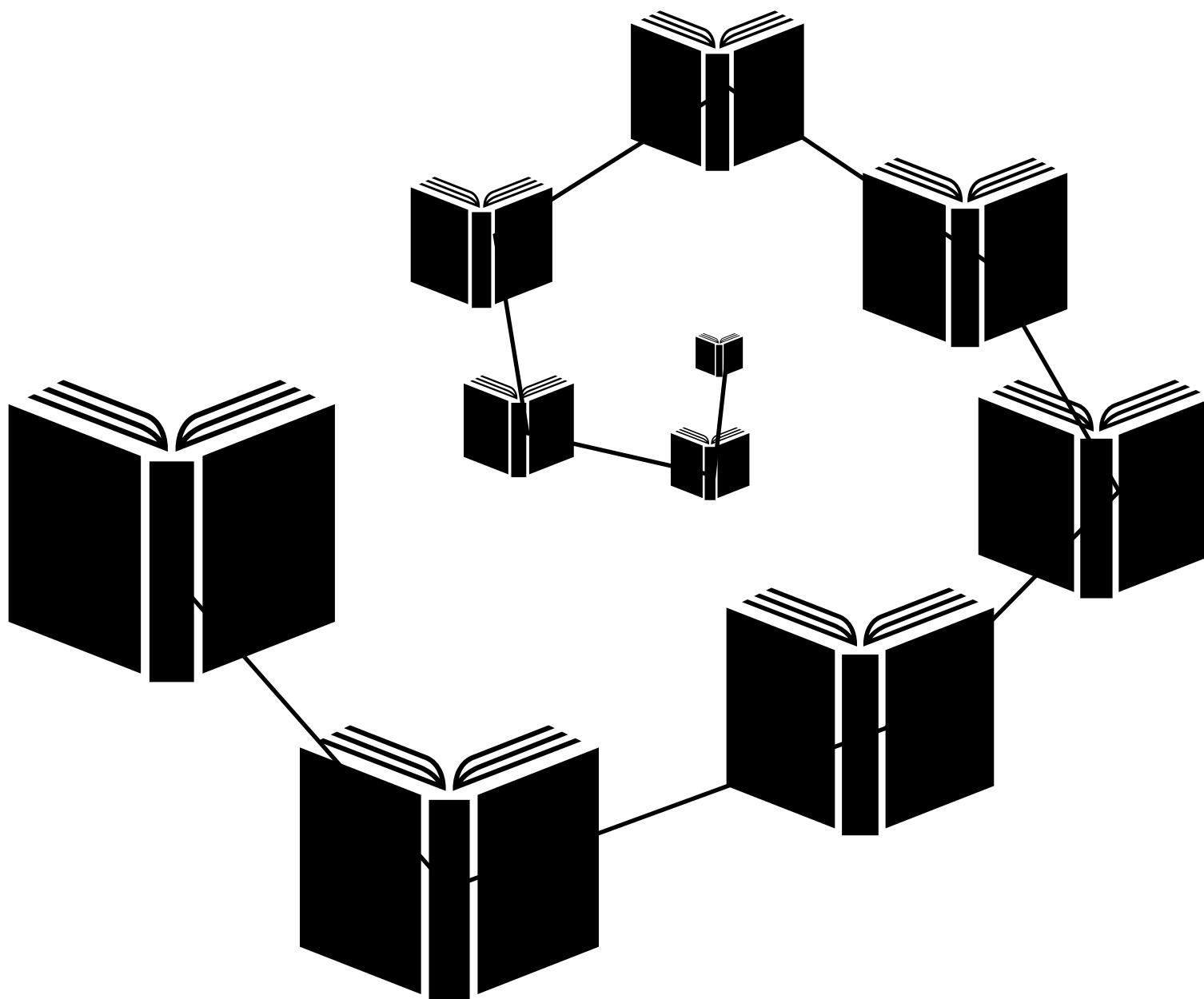
Active



Why Do Theory

Why Do Theory

Infinite **learning opportunities** of beautiful facts

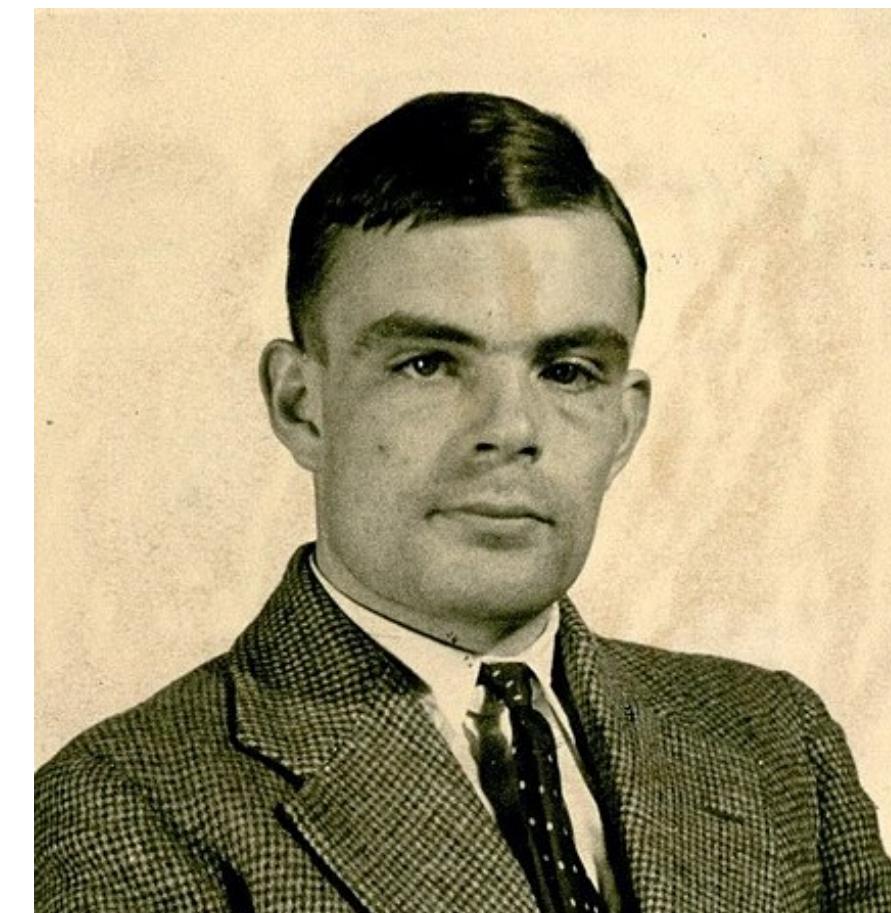


Why Do Theory

It's a **young** field (less to get up to speed with)



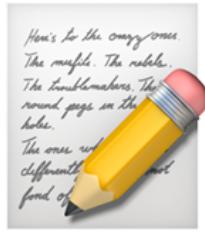
~2000 Years Ago



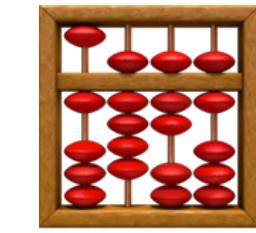
~100 Years Ago

Why Do Theory

Uniquely at the intersection of the **creative** and the **formal**



and (sometimes) the **practical**



Why Do Theory

Theory

+

“*Music is the only **magic** left in this world.*”



-Bob Dylan

-My dad

1. guided by arcane laws
2. results often defy common sense
3. takes intense study to master