

# **On How to Learn, Do and Write Theory**

**Spring 2026  
Brown University**

**D Ellis Hershkowitz**

# 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

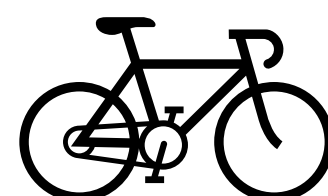


# How Theory is Done

**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.

...

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

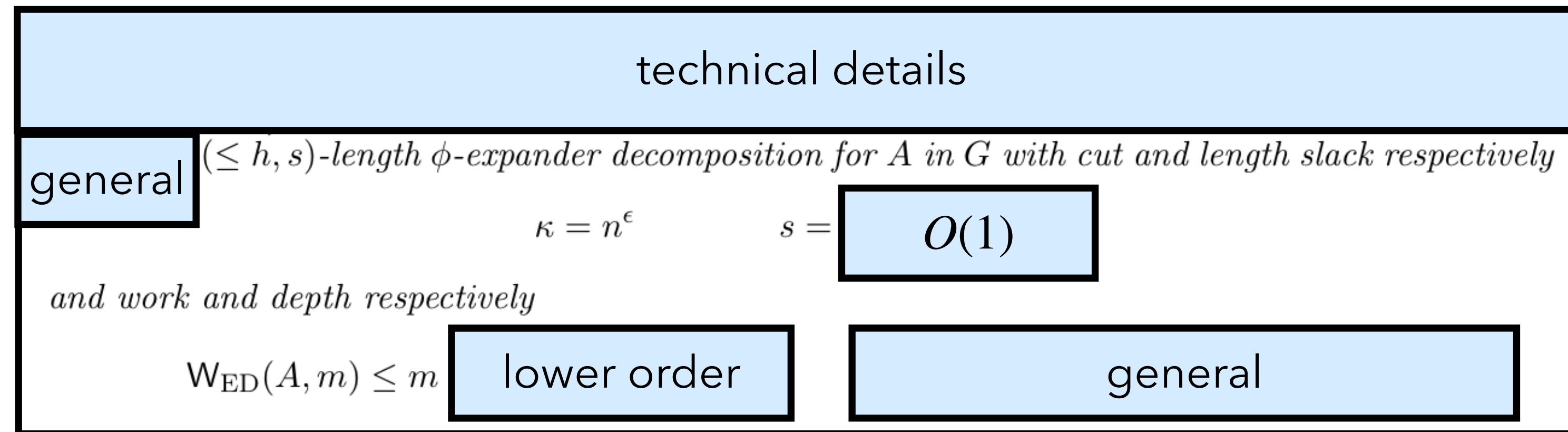
≈Terry Tao

# How to Learn Theory



# How to Learn Theory

## Simplification



*ignore lower order parameters / technical details*

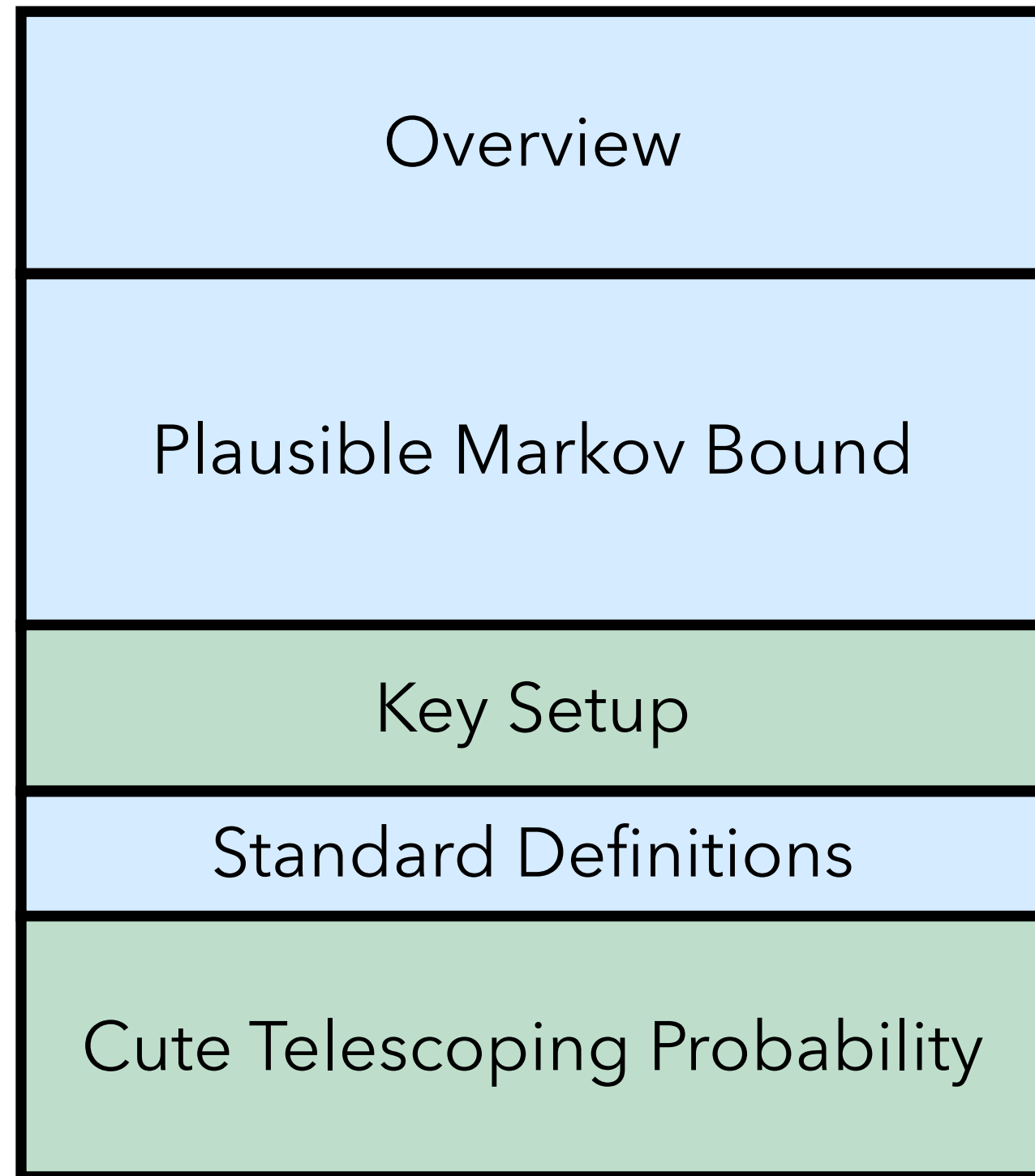
*fix parameters*

*apply theorem to special cases*

**Simplify** theorems

# How to Learn Theory

## Simplification

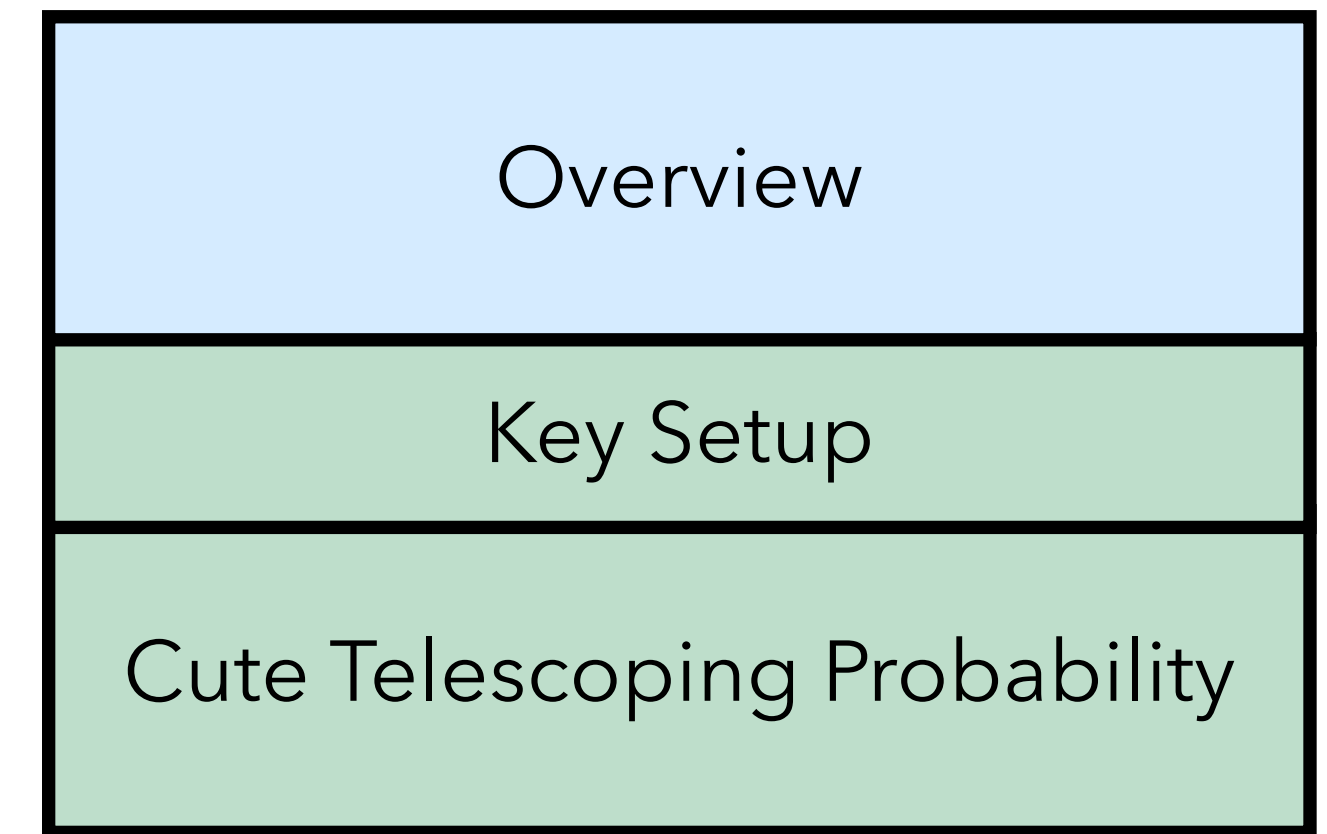


*Proof on Arbitrary Graphs*

*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  $\Pr[S_w = 1 | X_w = 1]$  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  $\pi$ , 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  $\pi$  is a random permutation, the probability that  $w_j$  comes before the  $x_k$  for all  $k < j$  is exactly  $1/j$ . Thus  $\Pr[S_{w_j} = 1 | X_{w_j} = 1] \leq 1/j$ . So by setting  $b_{w_j} = 1/j$ , we have proved the first part of the lemma.

5

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

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

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

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

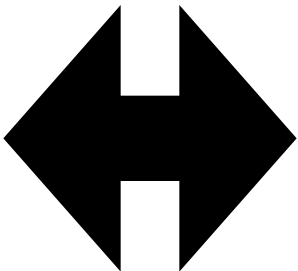
So we have that

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

Thus

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

where the final inequality is from the triangle inequality. ■



**2 Subclaims**

(A)  $\sum_i 2^{i+3} \Pr(X_w) \leq 16 d_{uw}$

(B)  $\exists$  bound  $b_w$  s.t.  $\Pr(S_w | X_w) \leq b_w$  and  $\sum b_w \leq O(\log n)$

**Proving (A)**

WLOG suppose  $d_{uw} \leq d_{vw} \Rightarrow$

Notice  $X_{uw}$  is 1 if  $u \in B(w, r_i)$  but  $v \notin B(w, r_i)$

$\Rightarrow \Pr(X_{uw})$  is  $\Pr(d_{uw} \leq r_i < d_{vw})$

Since  $r_i$  chosen uniformly in  $[2^{i-1}, 2^i)$  this is as  $\log d_{vw}$

Thus  $\sum_i 2^{i+3} \Pr(X_{uw}) = \sum_i \sum_{r_i \neq 0} 2^{i+3} \Pr(X_{uw}) \approx \sum_i 2^{i+3} \cdot \frac{\log d_{vw}}{\log d_{uw}} = 8 \left( \sum_{i=0}^{\log d_{uw}} 2^i - \sum_{i=0}^{\log d_{vw}} 2^i \right)$

$$= 8(2(d_{uw} - d_{vw})) = 16 d_{uw}$$

**Proving (B)**

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

$\pi$

Conditioning on a vert. cutting  $(u, v)$ , a vertex  $w$  settles  $(u, v)$  only if  $\Pi(w) < \Pi(u)$  &  $w$  is closer to  $(u, v)$  than  $u$

Thus if  $w$  is the  $j$ th closest vertex and we consider projecting  $\pi$  onto the  $j$  closest vertices then  $w$  must always come first  $\rightarrow$  halves w/  $\Pr \frac{1}{j}$

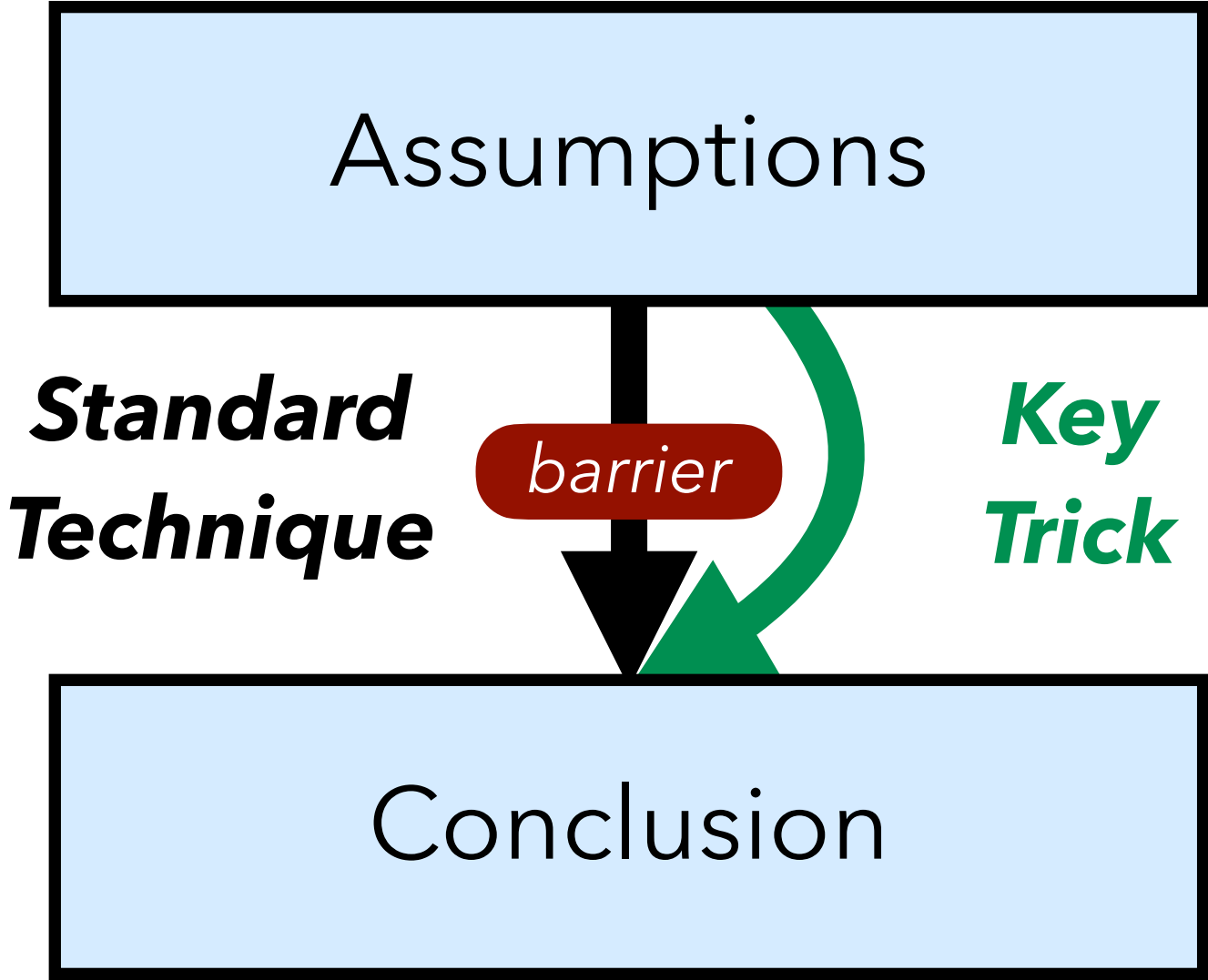
So let  $b_w = \frac{1}{j}$  and so  $\Pr(S_w | X_w) \leq \frac{1}{j}$

Sum of a harmonic gives  $\sum b_w \leq O(\log n)$

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

# How to Learn Theory

## Active Engagement



**Do the Same for LC Expander Decompositions?**

While  $G$  has an  $(h, s)$ -length  $\phi$ -sparse cut  $C$ :

Apply a length-constrained cut that is

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

50

**Problem 1: Union of Sparse LC Not Clearly Sparse**

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

51

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

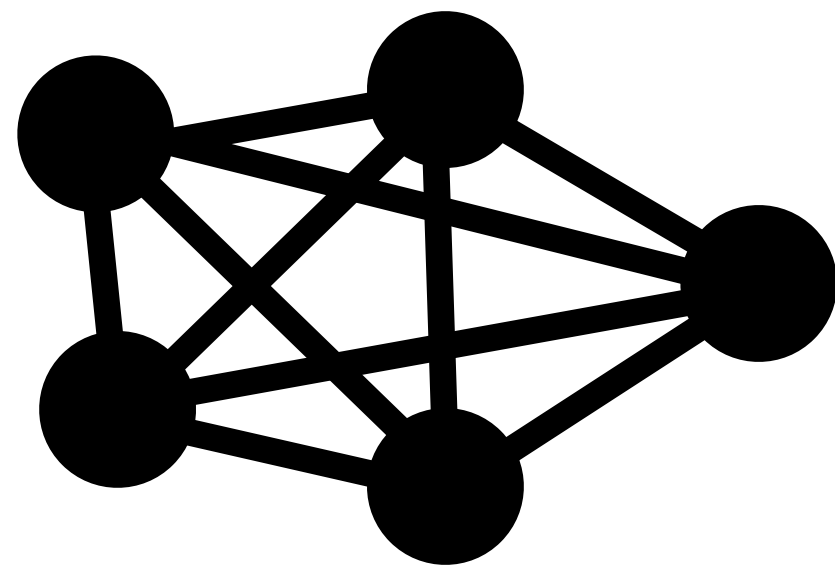
52

**Invent Stories** that you like / will remember

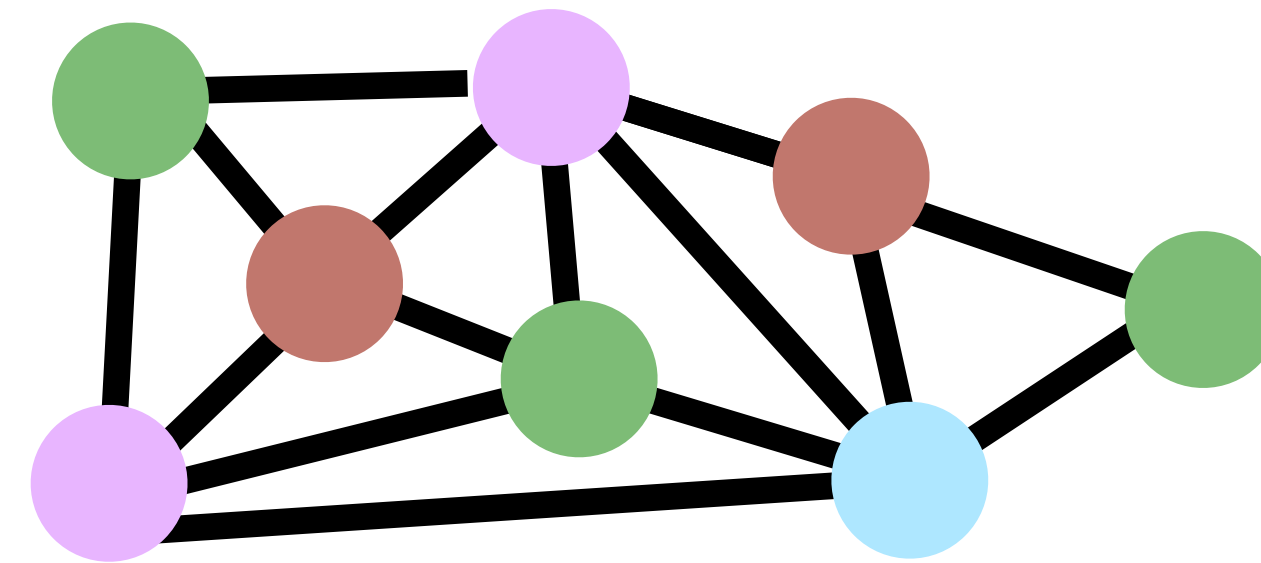
# How to Learn Theory

## Active Engagement

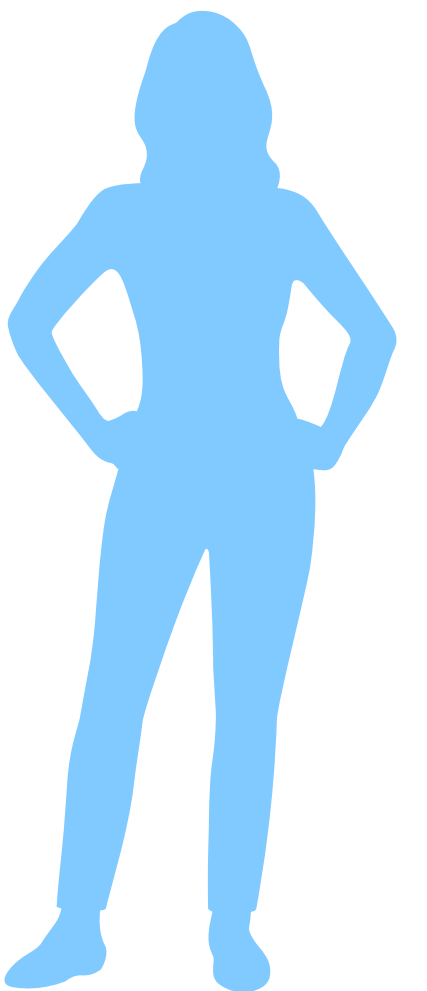
**Theorem:** Every planar graph is 4-colorable



*why assumptions needed?*



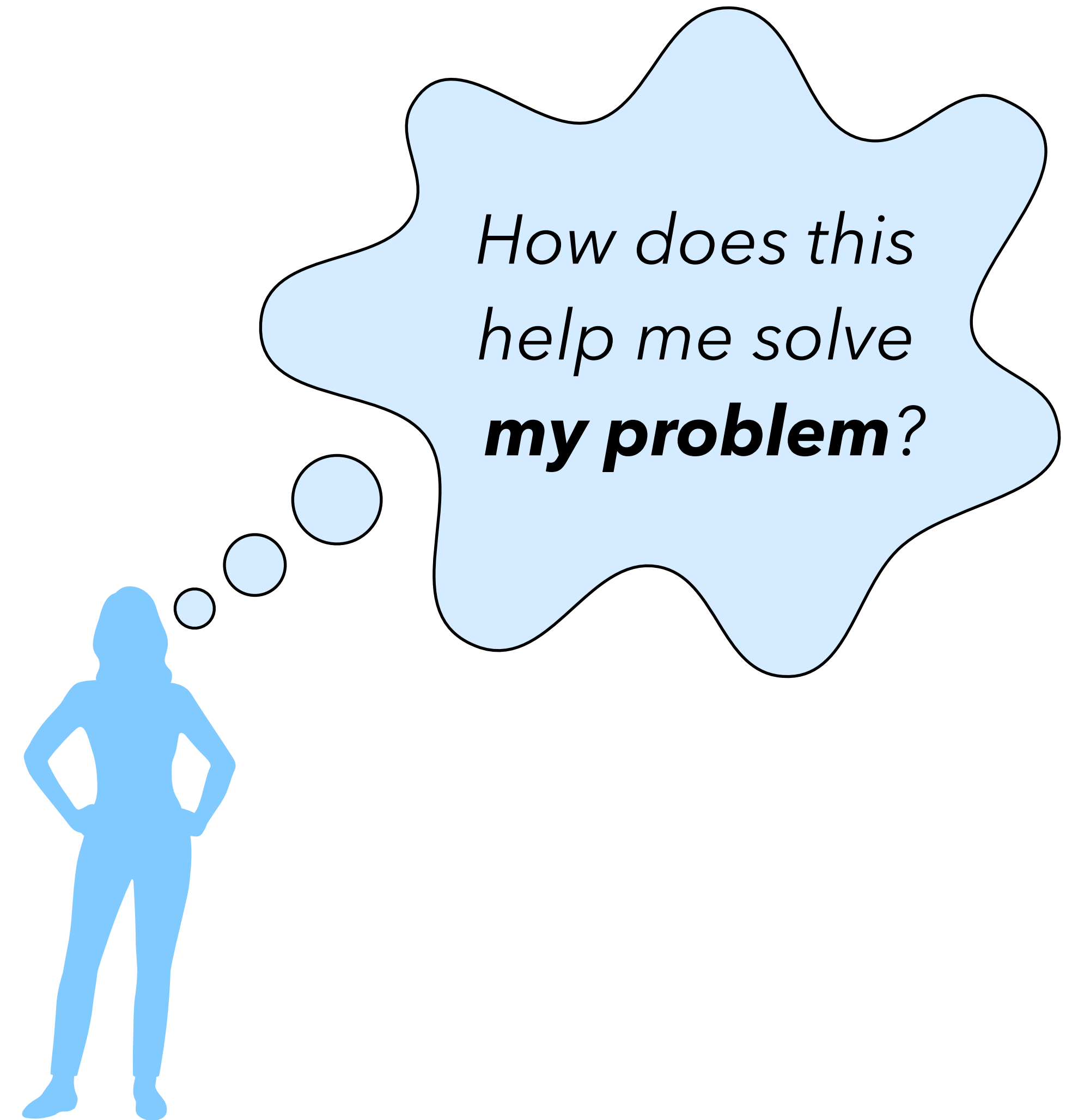
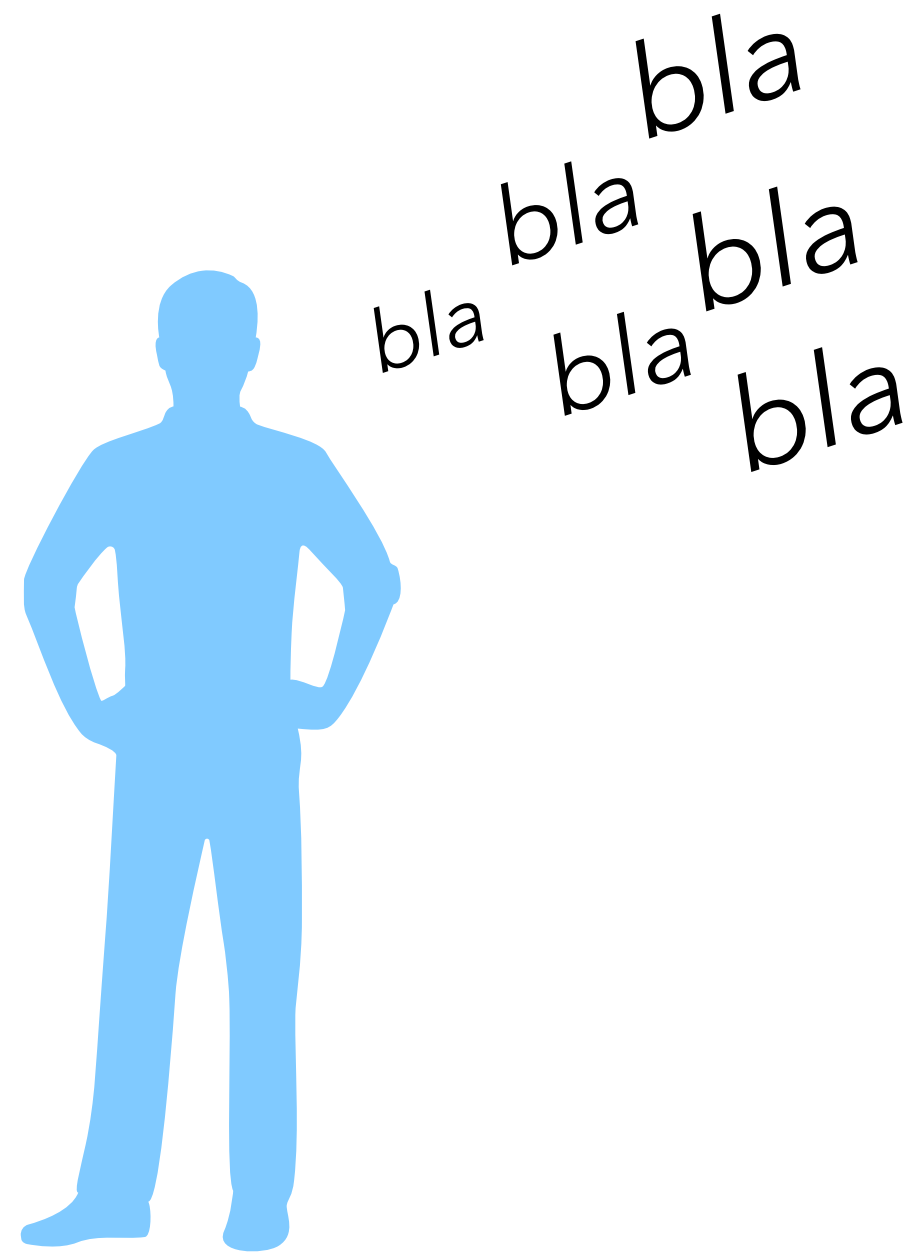
*what does this give on e.g.s?*



**Ask Yourself Questions**

# How to Learn Theory

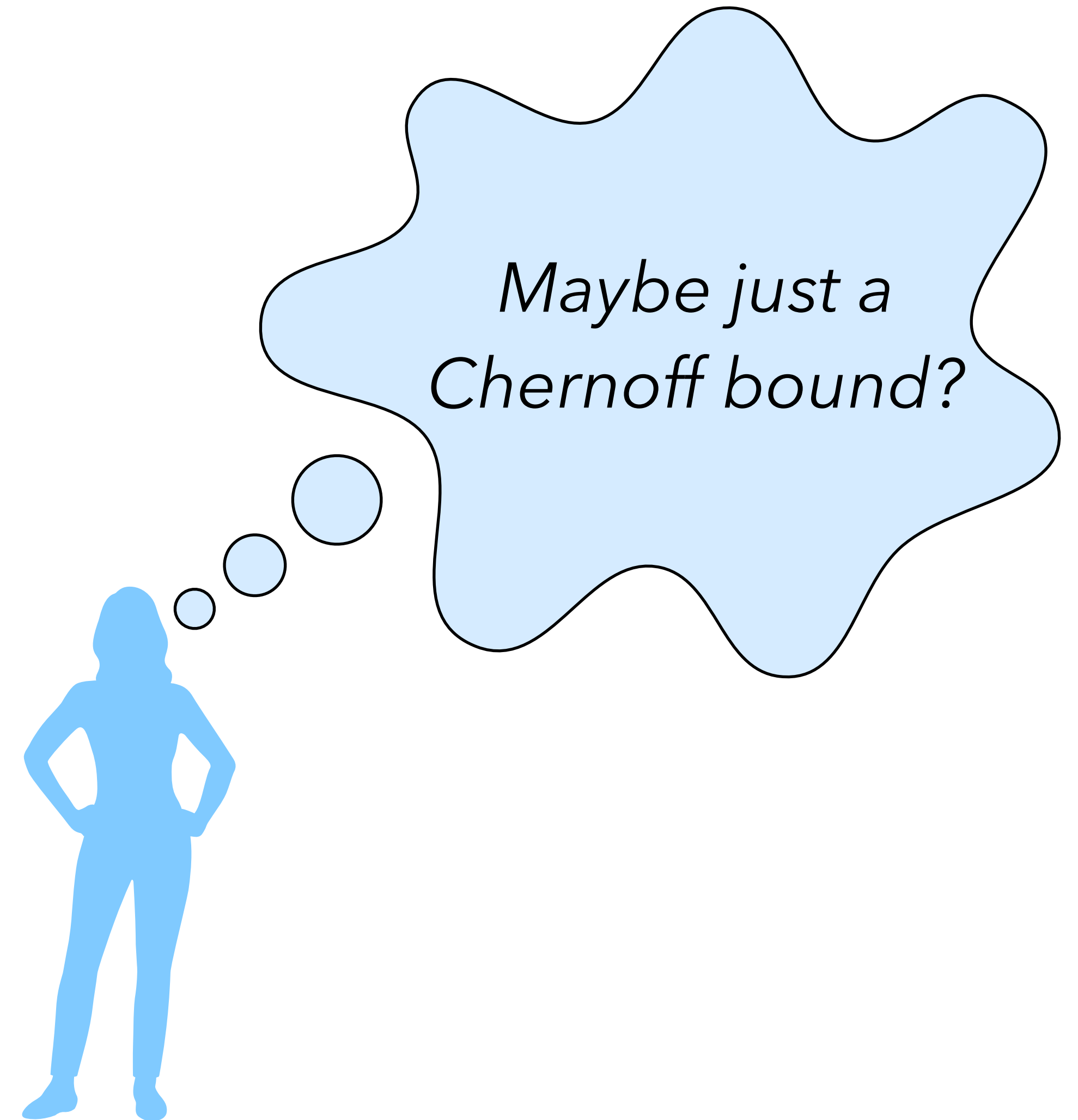
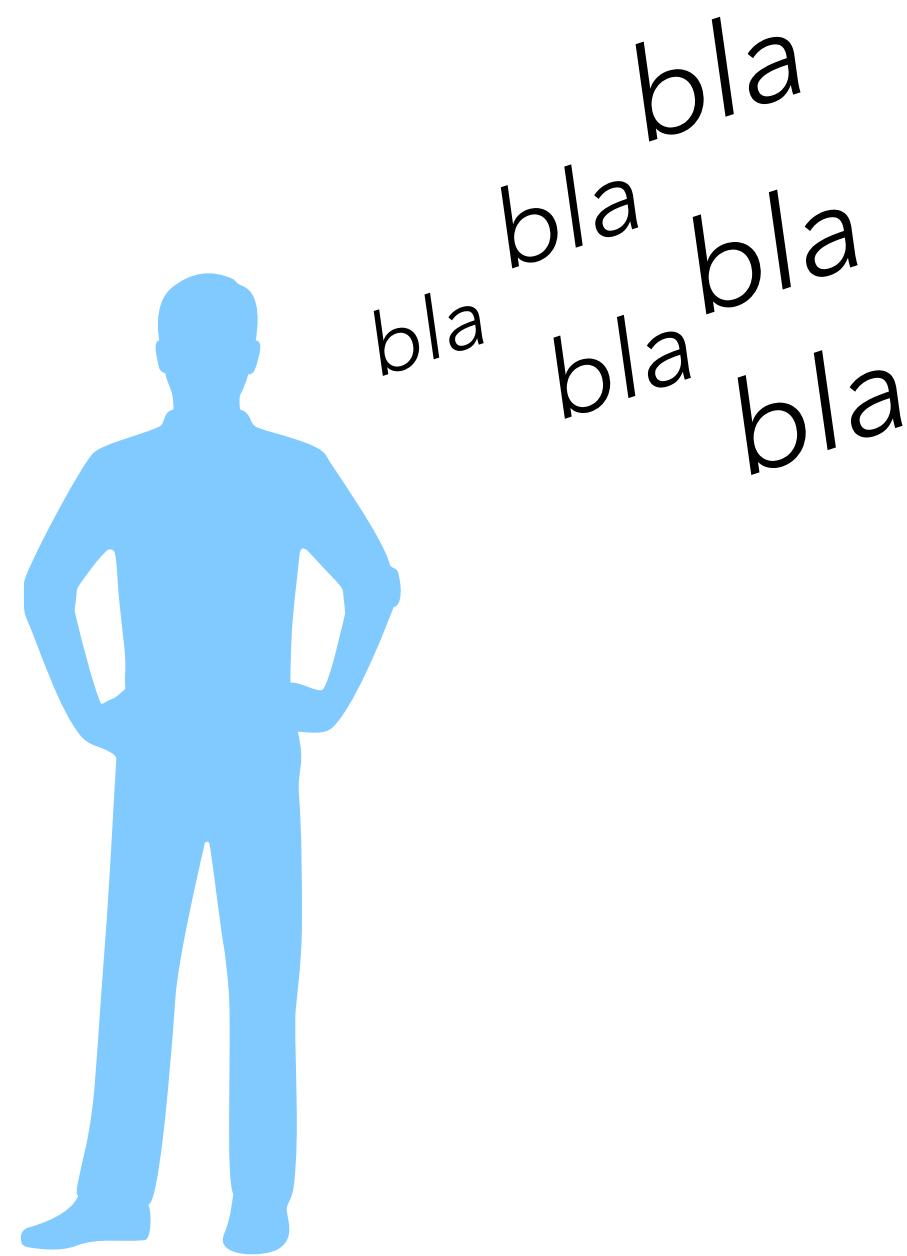
## Active Engagement



**Anchor Your Learning** with a problem you like

# How to Learn Theory

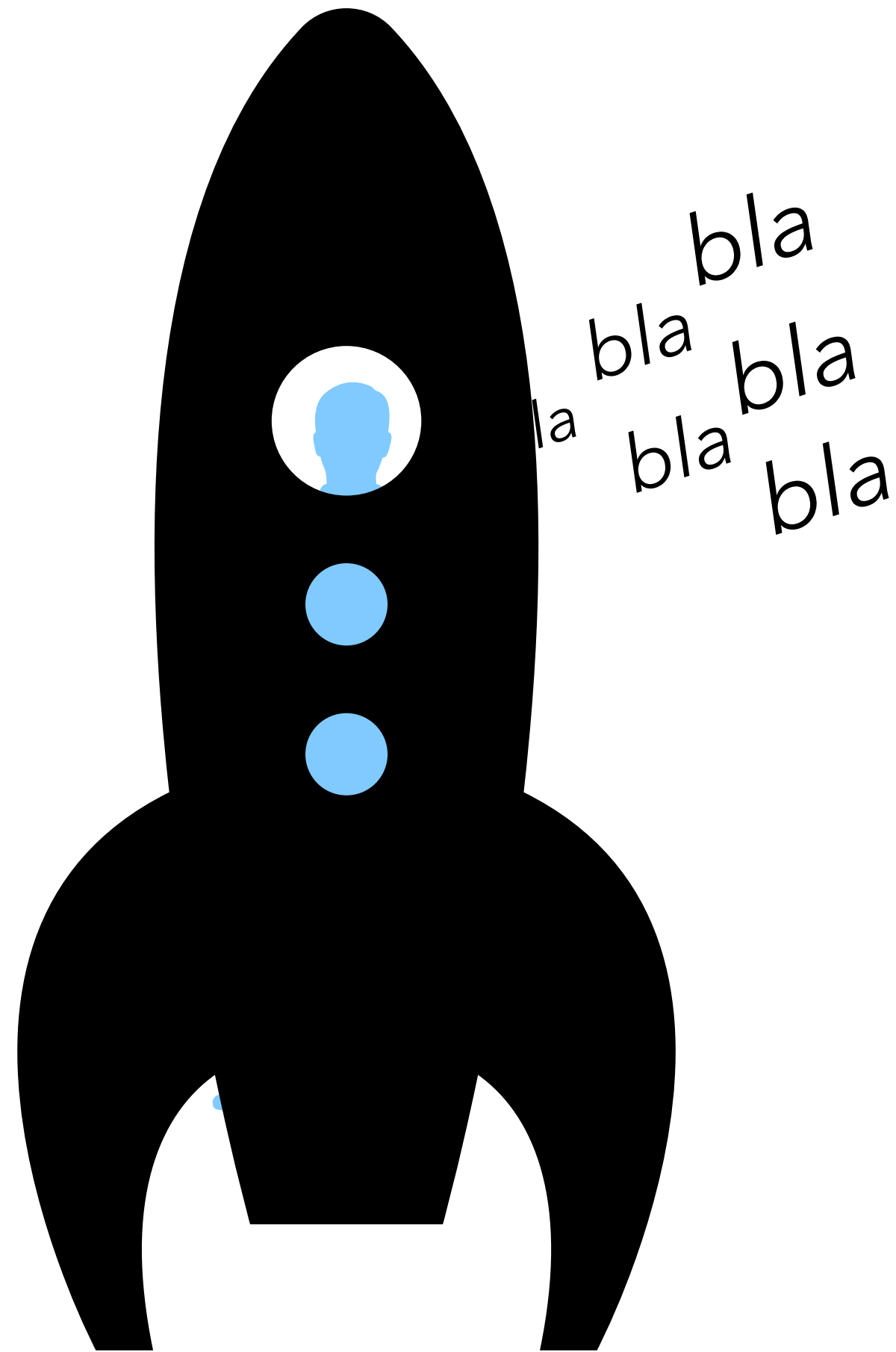
## Active Engagement



**Guess** what's coming next

# How to Learn Theory

## Active Engagement

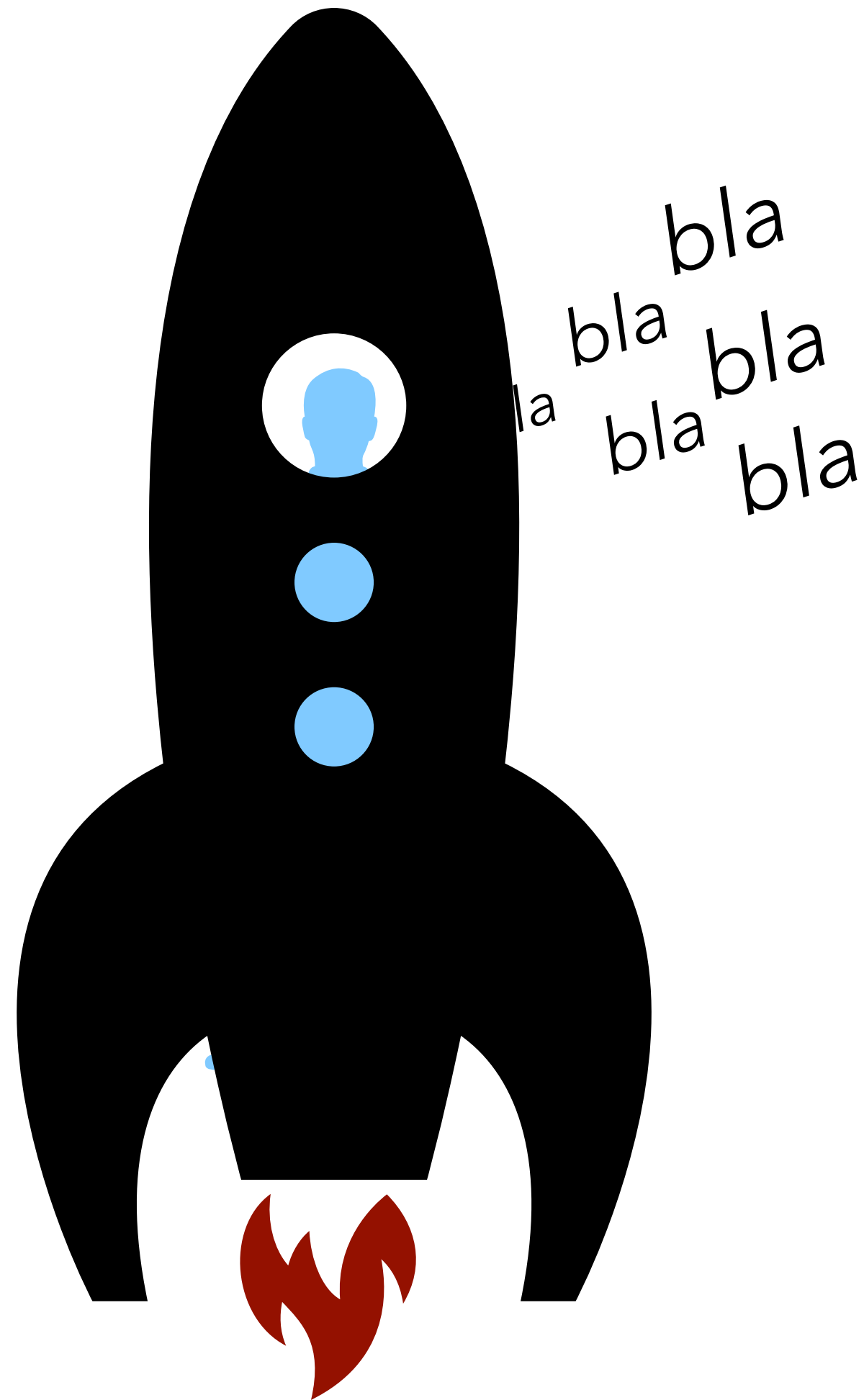


**Ask Questions** if you're confused



# How to Learn Theory

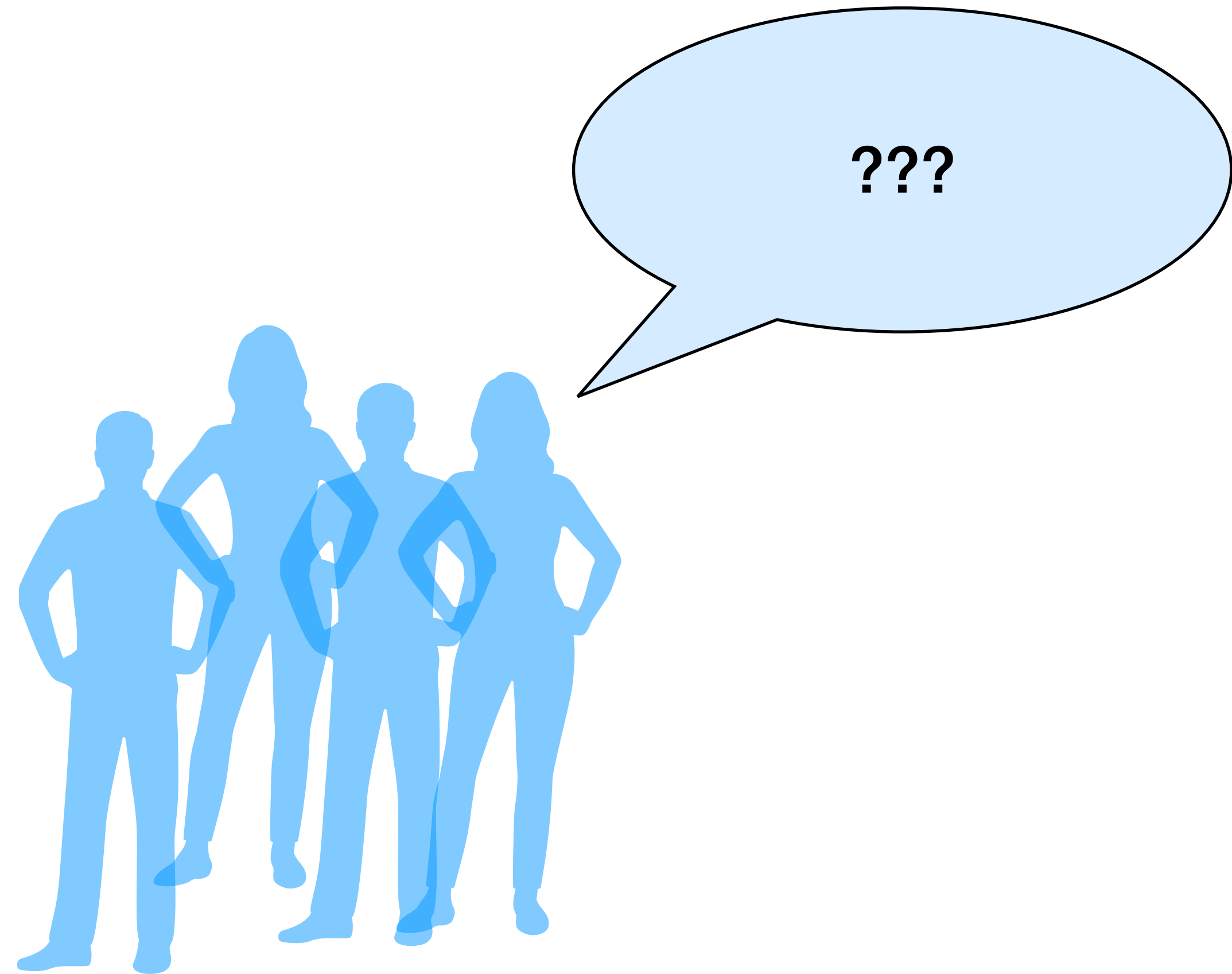
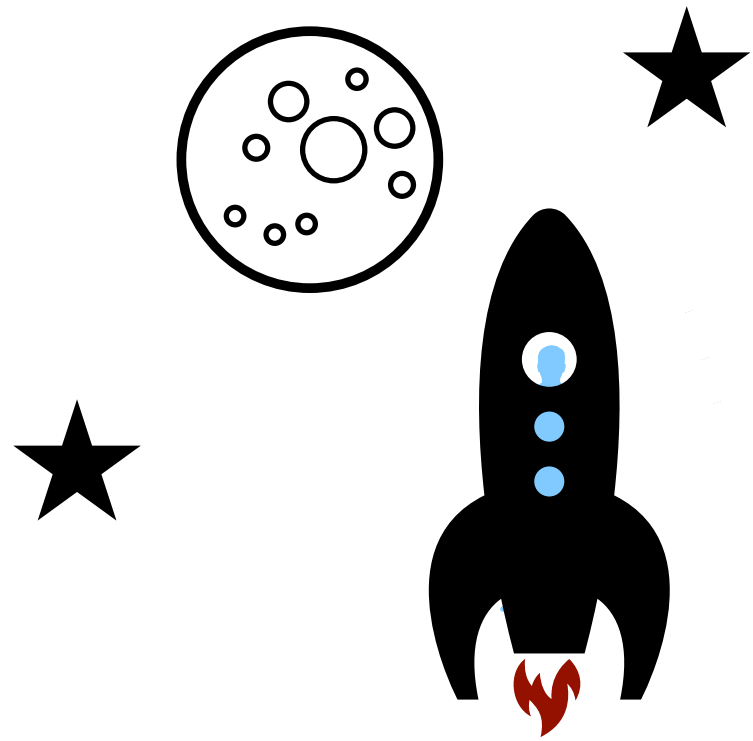
## Active Engagement



**Ask Questions** if you're confused

# How to Learn Theory

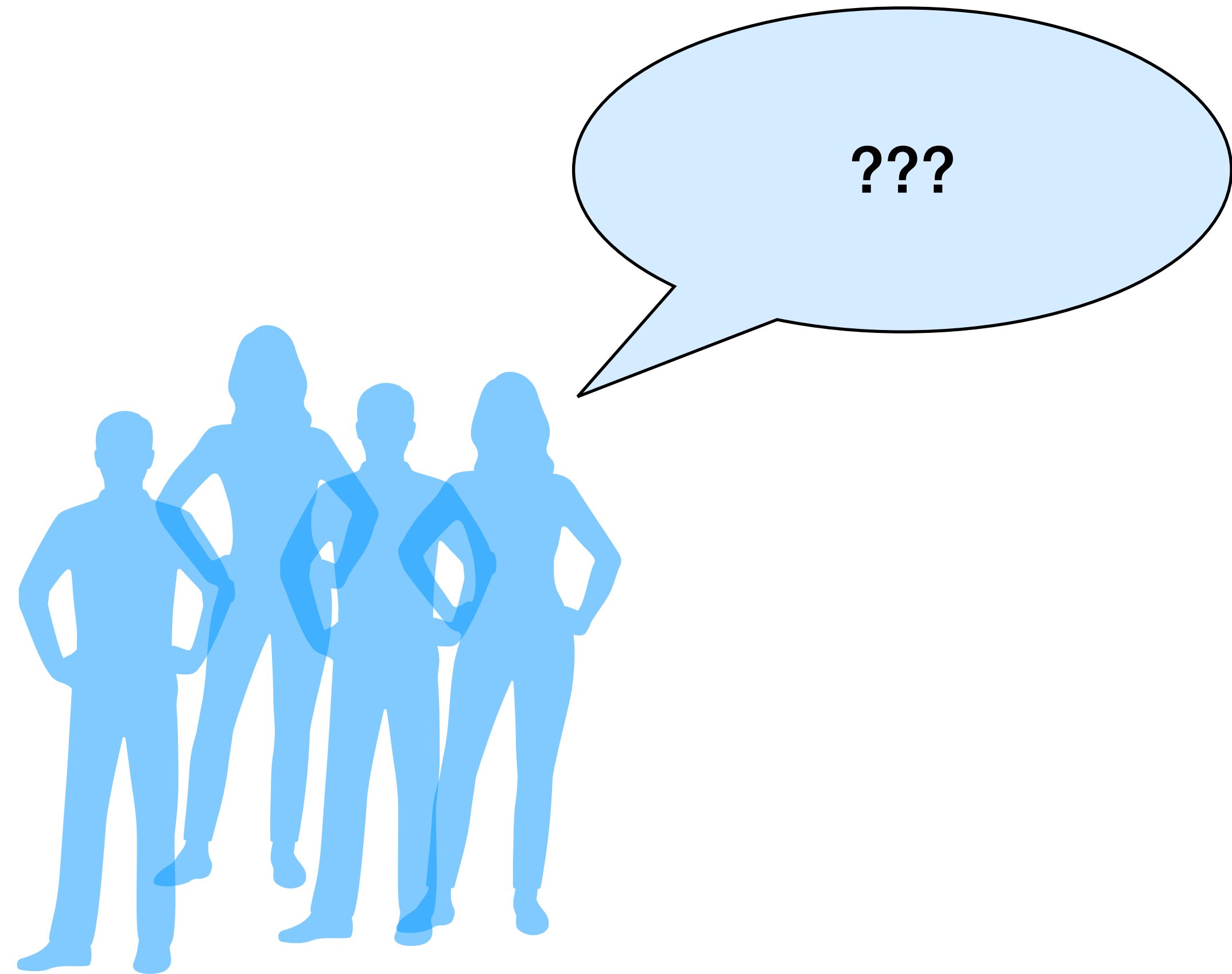
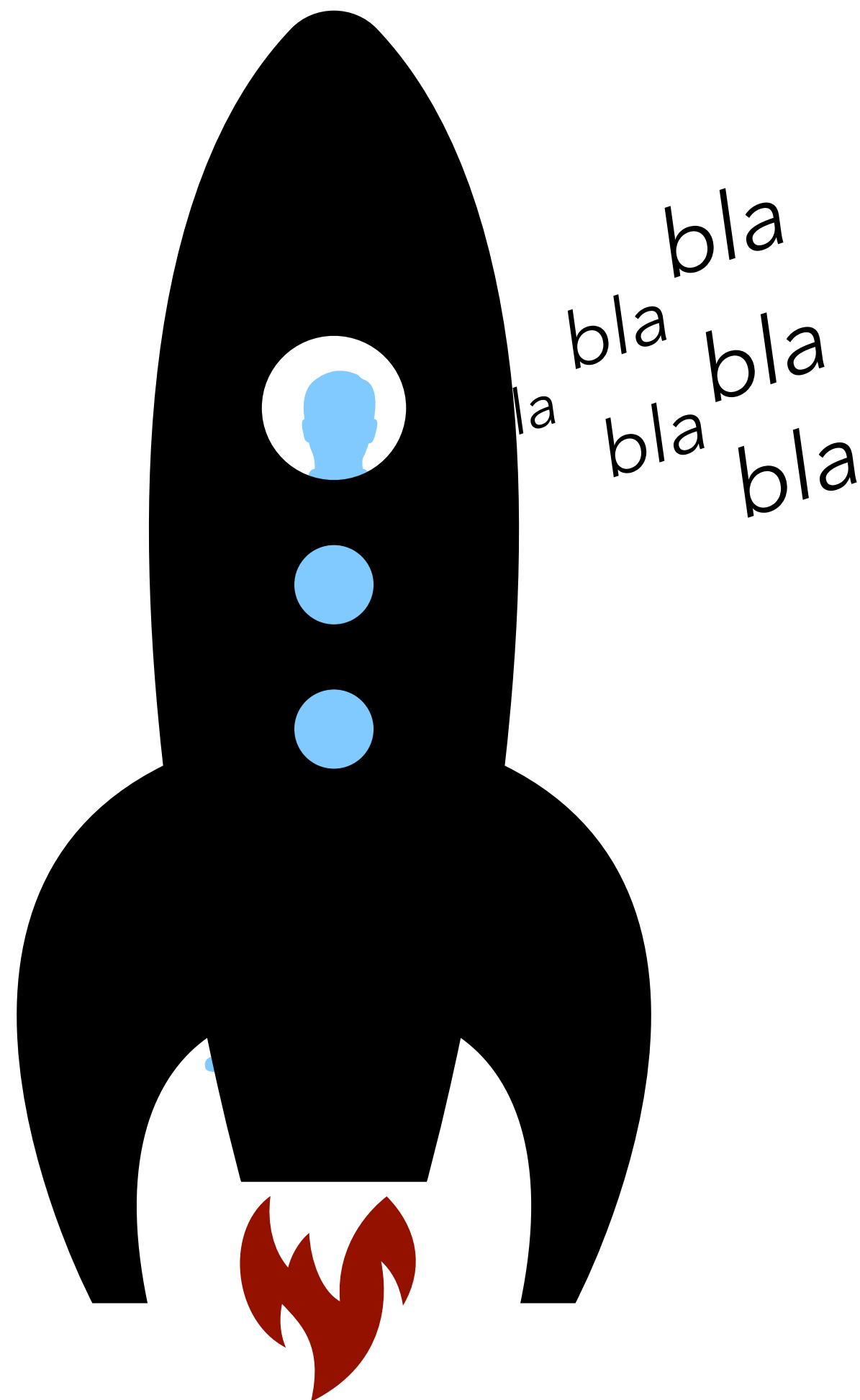
## Active Engagement



**Ask Questions** if you're confused

# How to Learn Theory

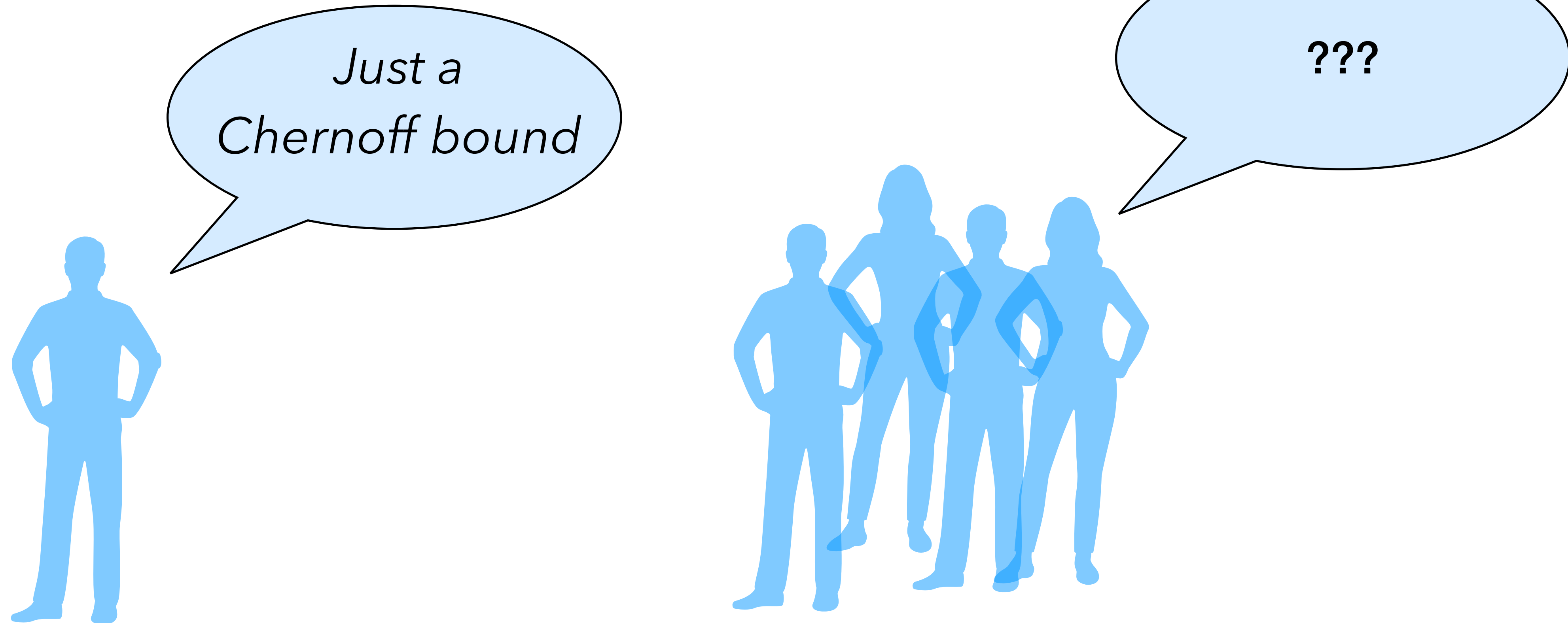
## Active Engagement



**Ask Questions** if you're confused

# How to Learn Theory

## Active Engagement



**Ask Questions** if you're confused

# How to Learn Theory

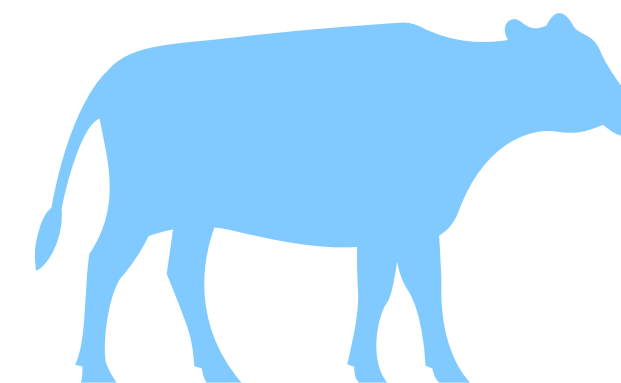
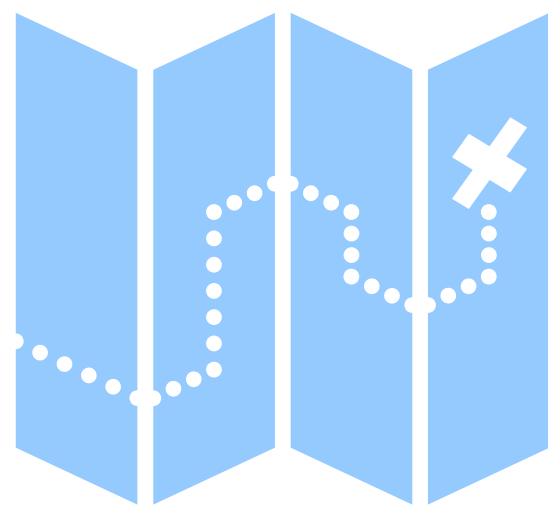
## What to Aim For

**Roadmaps** of proof

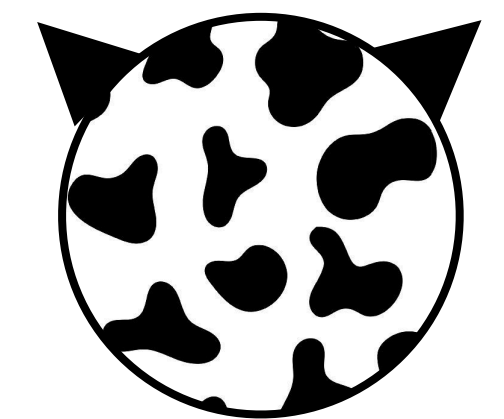
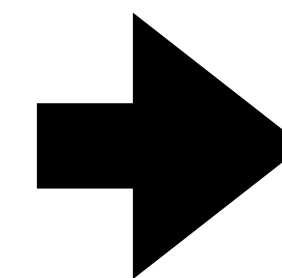


**Tools and stories** you'll remember

**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 **quickness** 
  - **Good memory** 
  - Good **intuition** 
  - Reliable
- Work to Your Strengths**
- Just really **curious** 
  - **Stick-To-Itiveness** 
  - **Impatient** / only interested in elegant solutions 
  - ...

# How to Do Theory

**You Will Get Stuck**

**Theorem:** Every planar graph is 4-colorable 🥵

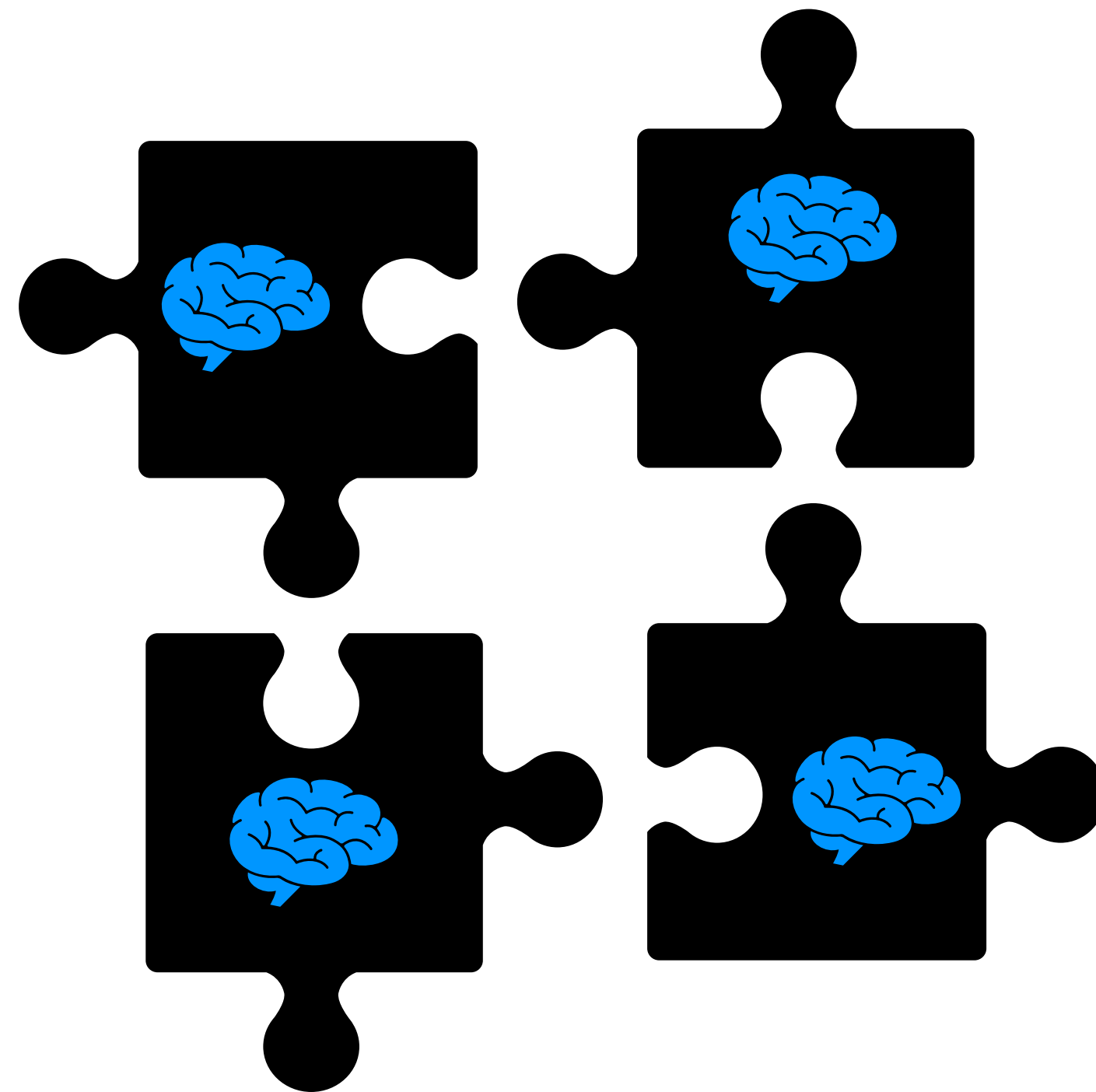
**Theorem:** Every tree is 4-colorable 😊

**Simplify your problem**



# How to Do Theory

**You Will Get Stuck**



**Collaborate**

# How to Do Theory

You Will Get Stuck

Collaborations

## STOC 2021

- “A (Slightly) Improved Approximation Algorithm for Metric TSP”, by Anna R. Karlin (University of Washington), Nathan Klein (University of Washington), and Shayan Oveis Gharan (University of Washington).
- “The Complexity of Gradient Descent:  $CLS = PPAD \cap PLS$ ”, by John Fearnley (University of Liverpool), Paul W. Goldberg (University of Oxford), Alexandros Hollender (University of Oxford), and Rahul Savani (University of Liverpool).
- “Indistinguishability Obfuscation from Well-Founded Assumptions”, by Aayush Jain (University of California at Los Angeles), Huijia Lin (University of Washington), and Amit Sahai (University of California at Los Angeles).

## STOC 2020

- “Improved Bounds for The Sunflower Lemma”, by Ryan Alweiss, Shachar Lovett, Kewen Wu, and Jiapeng Zhang.

## STOC 2019

- “Log-Concave Polynomials II: High-Dimensional Walks and an FPRAS for Counting Basis of a Matroid”, by Nima Anari, Kuikui Liu, Shayan Oveis Gharan, and Cynthia Vinzant.
- “The Reachability Problem for Petri Nets Is Not Elementary”, by Wojciech Czerwiński, Sławomir Lasota, Ranko Lazić, Jérôme Leroux, and Filip Mazowiecki.
- “Oracle Separation of BPQ and PH”, by Ran Raz and Avishay Tal.

## STOC 2018

- “A Constant-Factor Approximation Algorithm for the Asymmetric Traveling Salesman Problem”, by Ola Svensson, Jakub Tarnawski, and László A. Végh.

## STOC 2017

- “Explicit, Almost Optimal, Epsilon-Balanced Codes”, by Amnon Ta-Shma.
- “Deciding Parity Games in Quasipolynomial Time”, by Cristian Calude, Sanjay Jain, Bakhadyr Khoussainov, Wei Li, and Frank Stephan.
- “A Weighted Linear Matroid Parity Algorithm”, by Satoru Iwata and Yusuke Kobayashi.

## STOC 2016

- “Reed-Muller Codes Achieve Capacity on Erasure Channels”, by Shrinivas Kudekar, Santhosh Kumar, Marco Mondelli, Henry D. Pfister, Eren Sasoglu, and Rudiger Urbanke.
- “Explicit Two-Source Extractors and Resilient Functions”, by Eshan Chattopadhyay and David Zuckerman.
- “Graph Isomorphism in Quasipolynomial Time”, by László Babai.

## STOC 2015

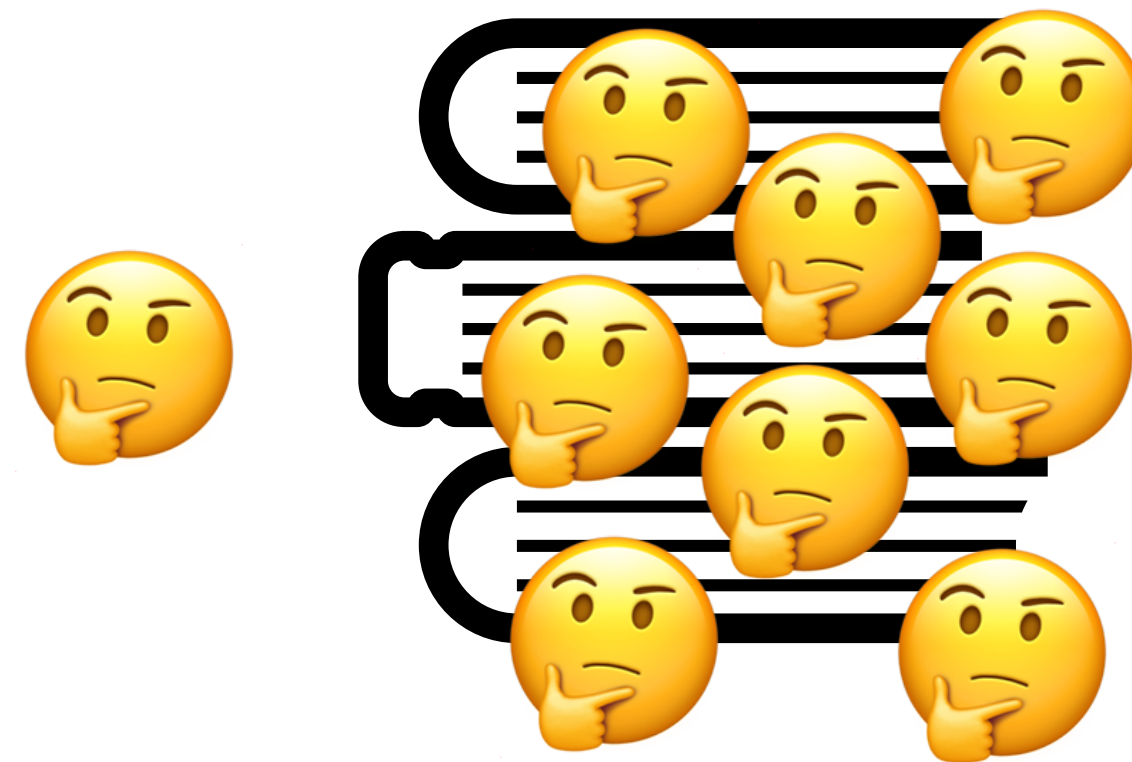
- “Exponential Separation of Information and Communication for Boolean Functions”, by Anat Ganor, Gillat Kol, and Ran Raz.
- “Lower Bounds on the Size of Semidefinite Programming Relaxations”, by James Lee, Prasad Raghavendra, and David Steurer.
- “2-Server PIR with Sub-Polynomial Communication”, by Zeev Dvir and Sivakanth Gopi.

*STOC Best Papers*

Collaborate

# How to Do Theory

**You Will Get Stuck**



**Read Related Work**

# How to Do Theory

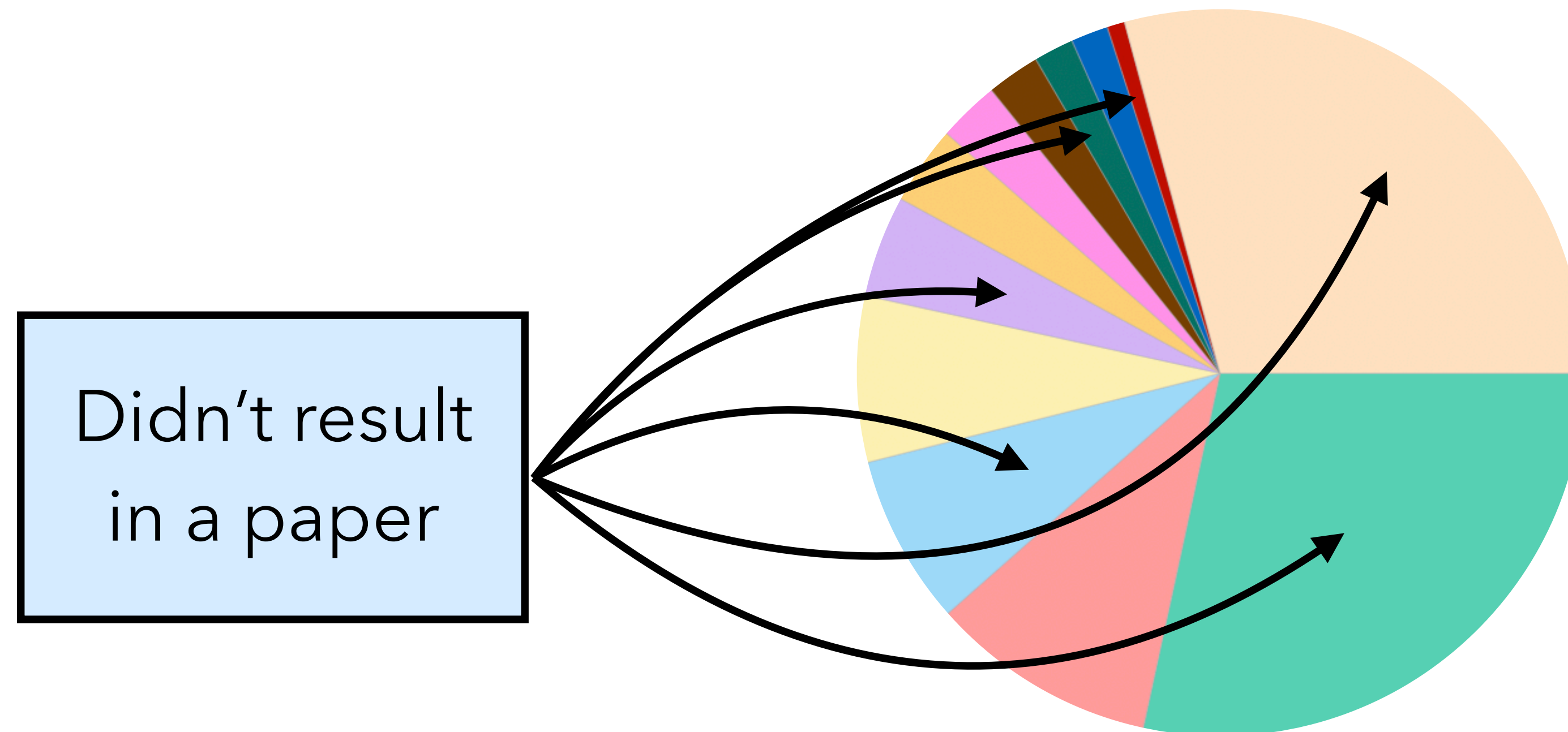
**You Will Get Stuck**



**Cut Yourself Slack**

# How to Do Theory

## A Few Mantras

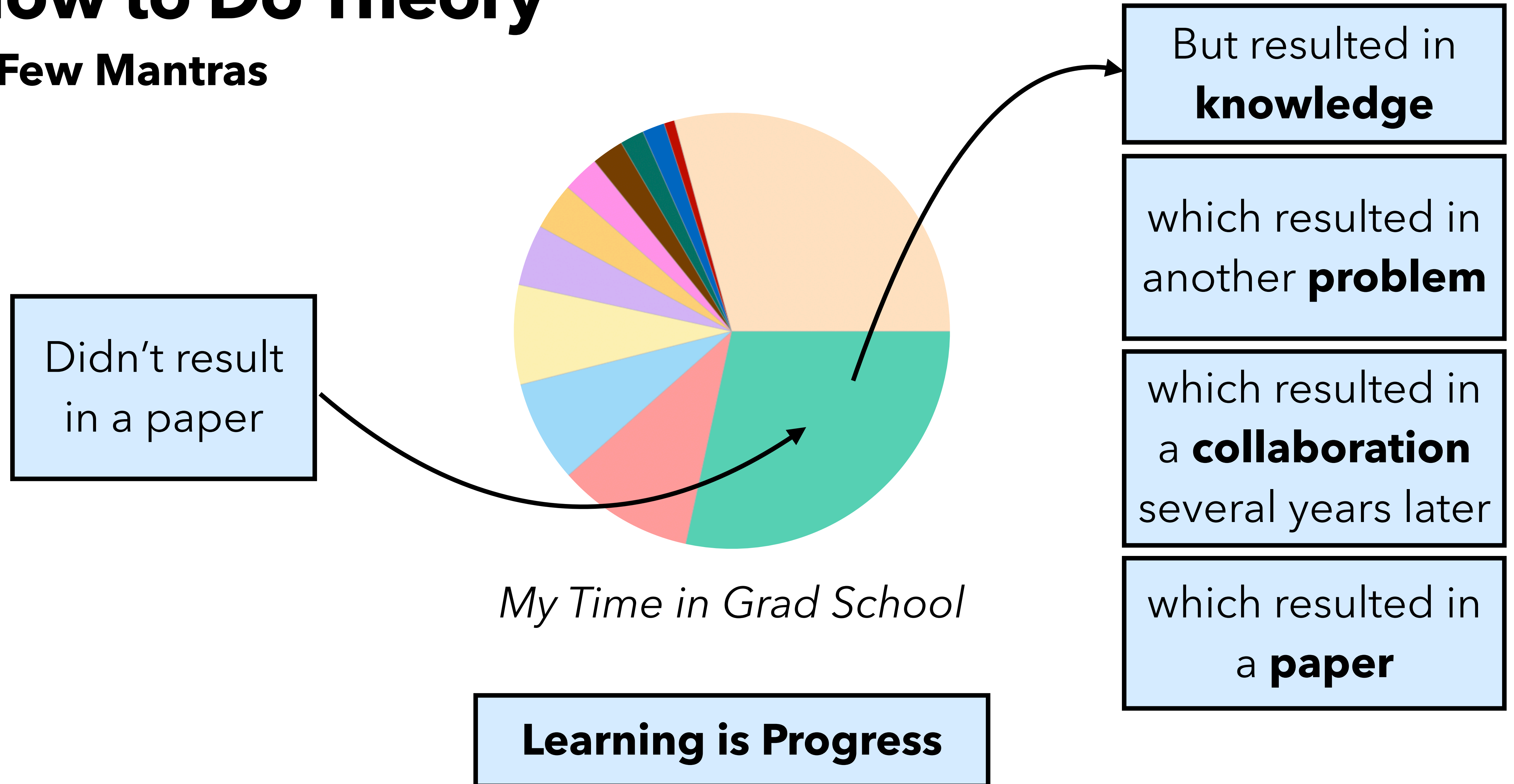


*My Time in Grad School*

**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**.
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.

...

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

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

**Proof:**

First we show  $1+1=2\dots$   
Next, we show  $2+2=4\dots$



**Lemma:**  $1+1=2$

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

**Proof:**

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

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

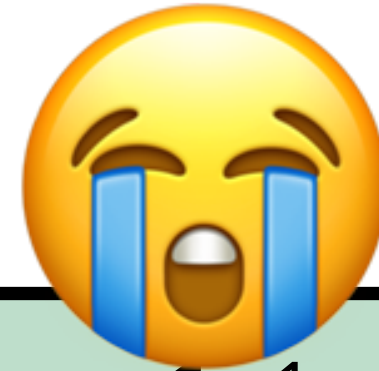
**Proof:**

By Lemma  $1+1=2$   
Next, we show  $2+2=4\dots$

Abstract out reused arguments into lemmas

# Writing Dos and Don'ts

## Bad References



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

**Proof:**

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

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

**Proof:**

By the argument in Theorem 1,  $1+1=2$   
Next, we show  $2+2=4\dots$



**Lemma:**  $1+1=2$

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

**Proof:**

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

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

**Proof:**

By Lemma  $1+1=2\dots$   
Next, we show  $2+2=4\dots$

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



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

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

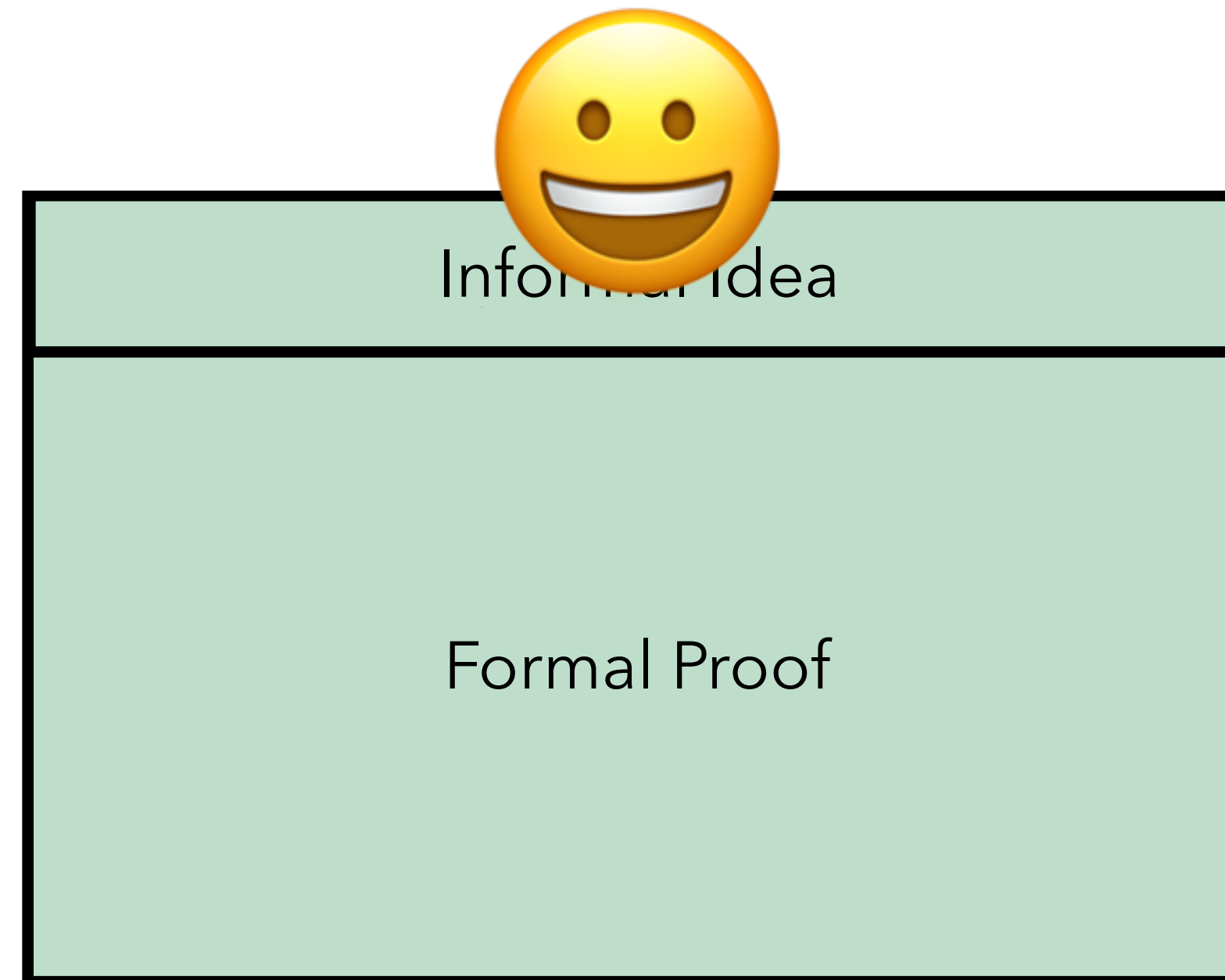
**Proof:**

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

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

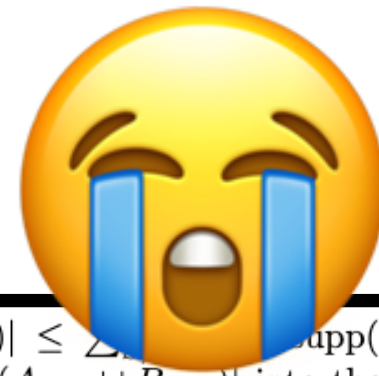
Just Right

Excruciating  
Formality

Balance intuition and formality

# Writing Dos and Don'ts

## General Style



and so  $\sum_{S,i} \sum_k |\text{supp}(A_{S,i,k} \cup B_{S,i,k})| \leq \sum_{S,i} |\text{supp}(A_S^{(i)})| \leq \tilde{O}(n^{1+O(\epsilon)} + N^{O(\epsilon)} L^2)$  Thus, plugging this bound on  $\sum_{S,i,k} |\text{supp}(A_{S,i,k} \cup B_{S,i,k})|$  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

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

Thus, plugging this bound on  $\sum_{S,i,k} |\text{supp}(A_{S,i,k} \cup B_{S,i,k})|$  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

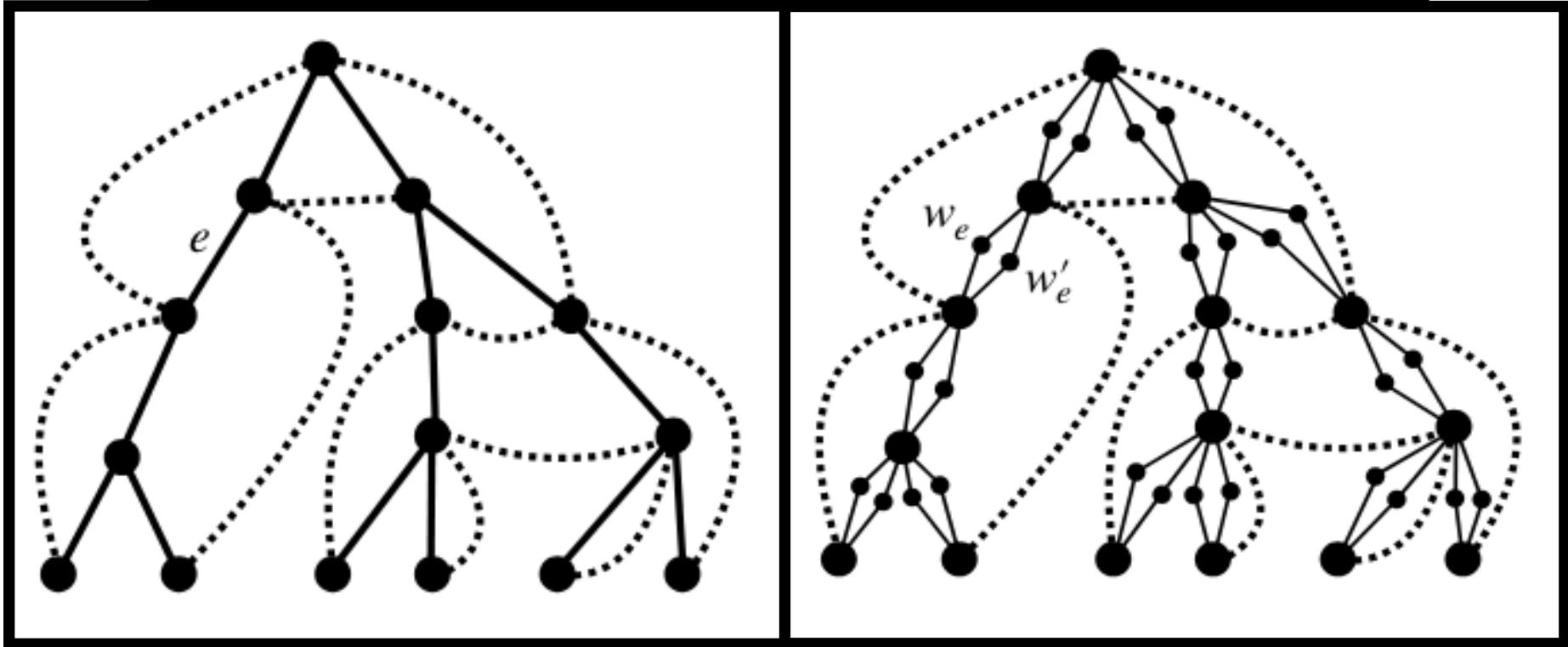
## General Style



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  $E_{\text{Gadget}}$  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  $E_{\text{Gadget}}$  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  $< x, y >$ .



Inner product  $\backslash\textbf{langle} x,y \textbf{rangle}\$$ .

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)



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

$ALG(x) = \log n$



$\$\texttt{\textbf{ALG}}(x) = \log n\$,$

$ALG(x) = \log n$

# 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.

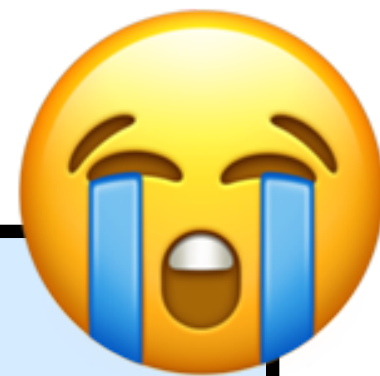


Let  $\$G\$$  be a  $\$k\$$ -connected graph.

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}.
```

We have

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

so  $A \leq C$  by Equation 1.



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

We have

$$\begin{array}{ll} A \leq B & \\ \leq C & (1) \end{array}$$

so  $A \leq C$  by Equation 1.





# 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.  $\backslash$  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**

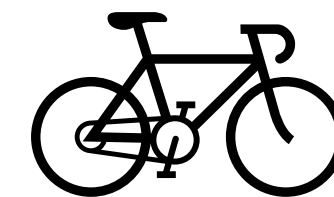
**Doing**

**Writing**

Simplification



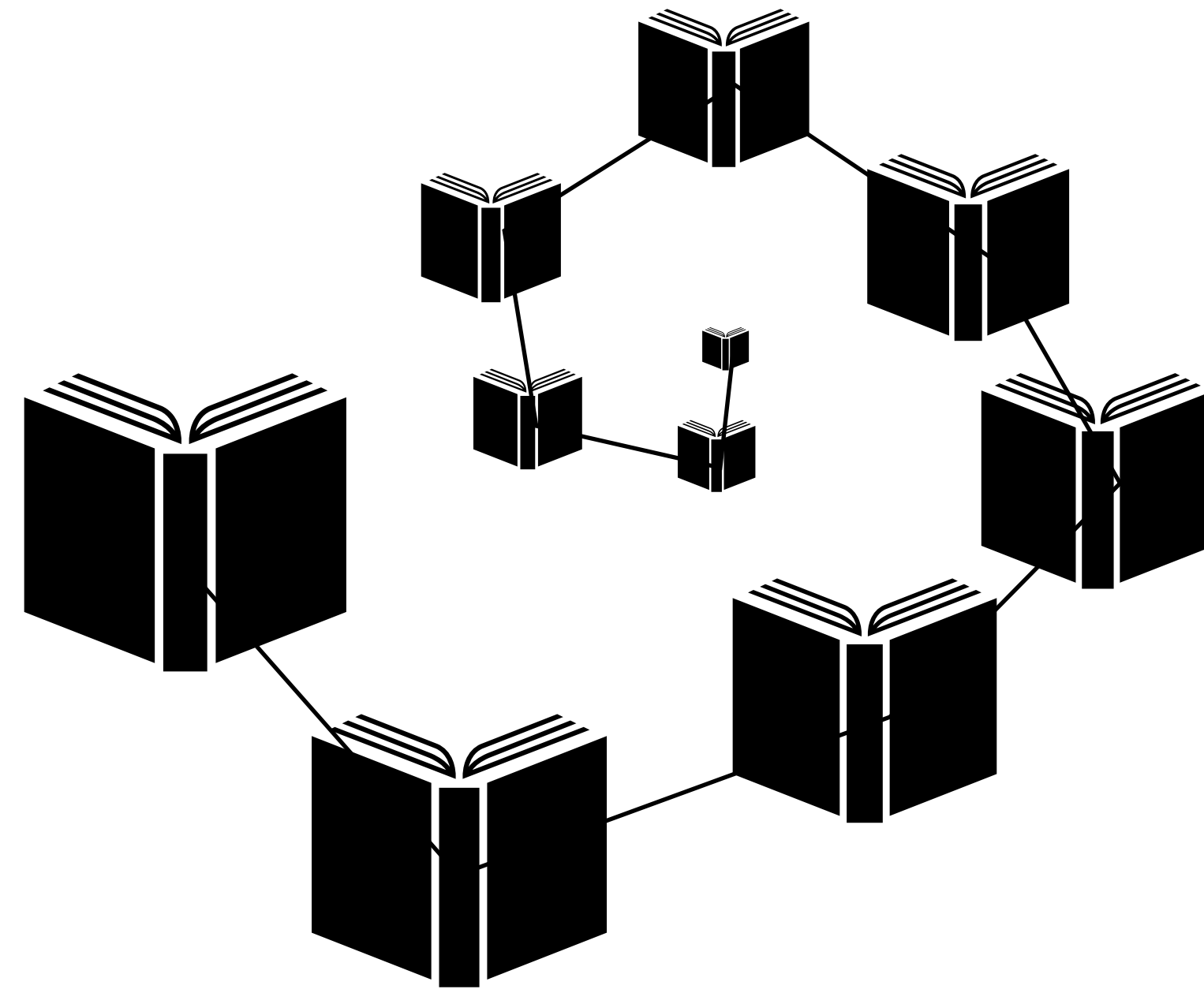
Active



# 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

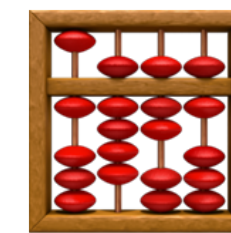


# Why Do Theory

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