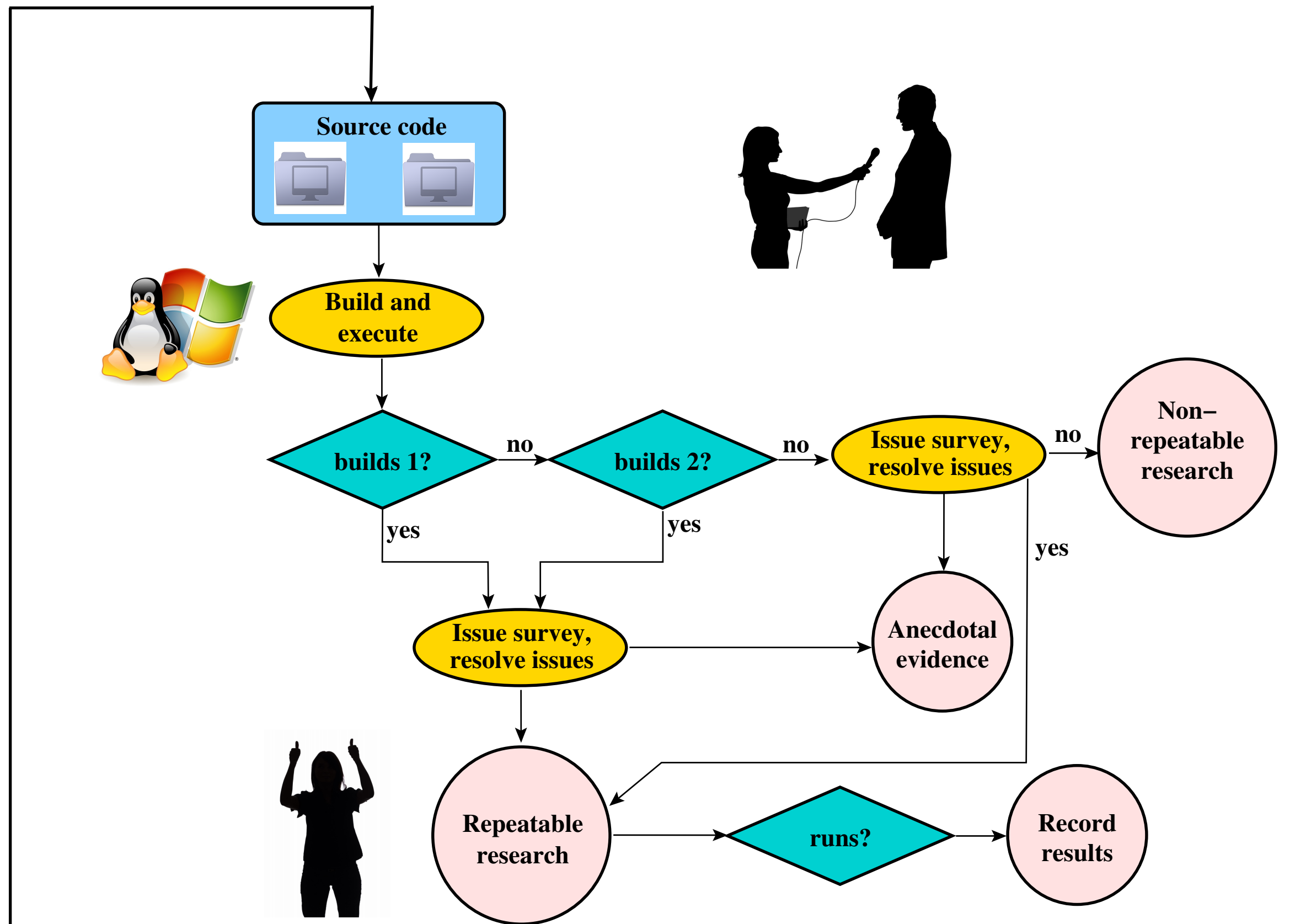
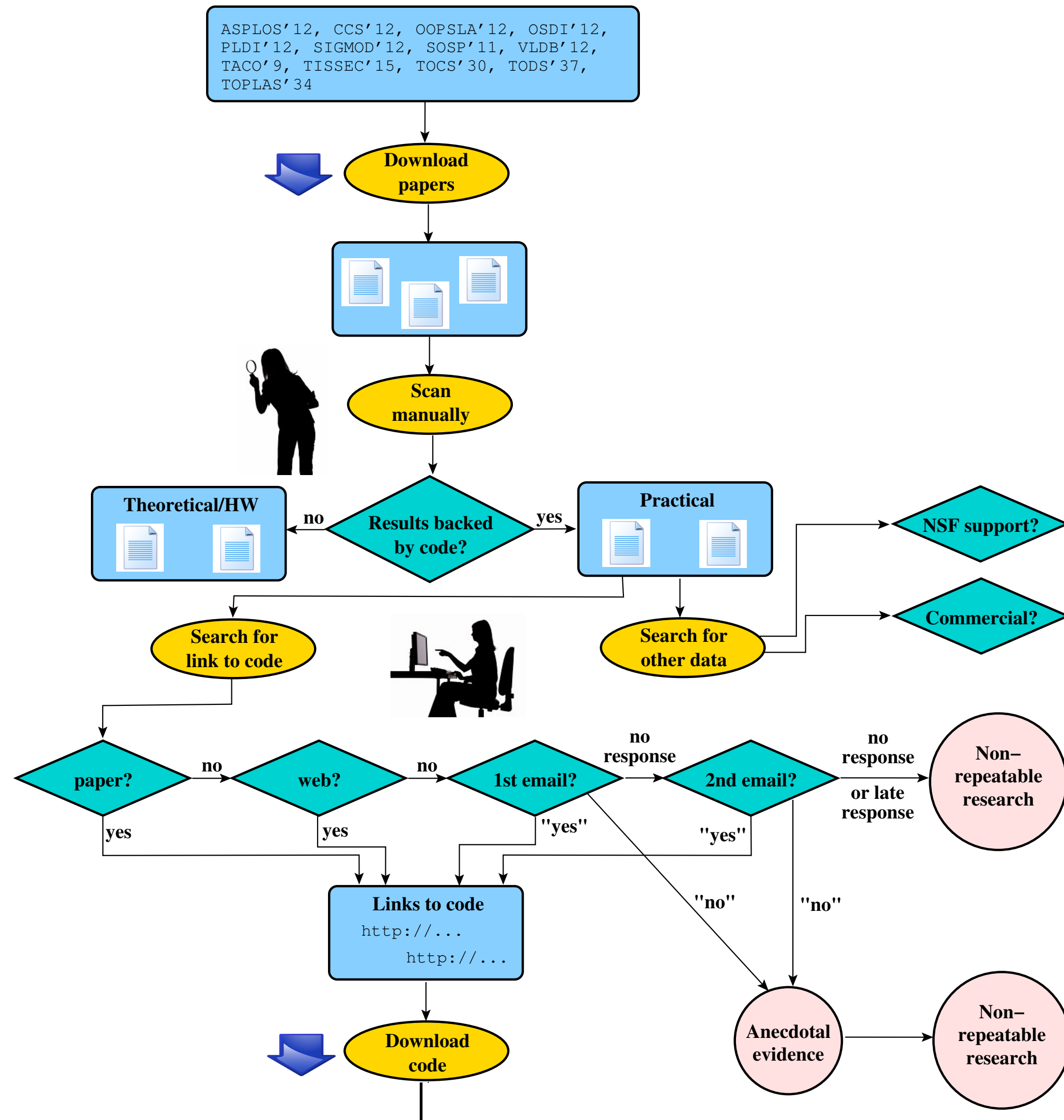


Advanced Data Management (CSCI 490/680)

Machine Learning in Databases

Dr. David Koop

Checking Computational Results in Systems



[Collberg and Proebsting, 2015]

Repeatability Results

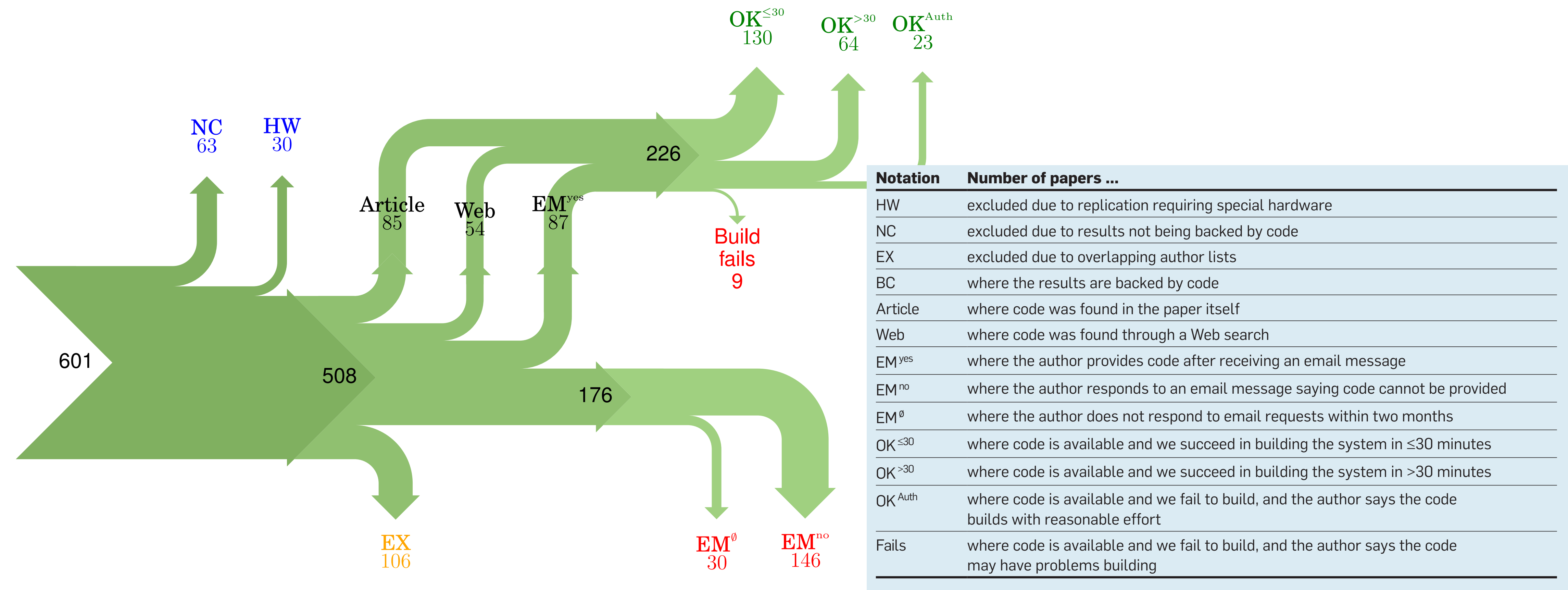


Figure 11: Study result. Blue numbers represent papers that were excluded from consideration, green numbers papers that are weakly repeatable, red numbers papers that are non-weakly repeatable, and orange numbers represent papers that were excluded (due to our restriction of sending at most one email to each author).

[Collberg and Proebsting, 2015]

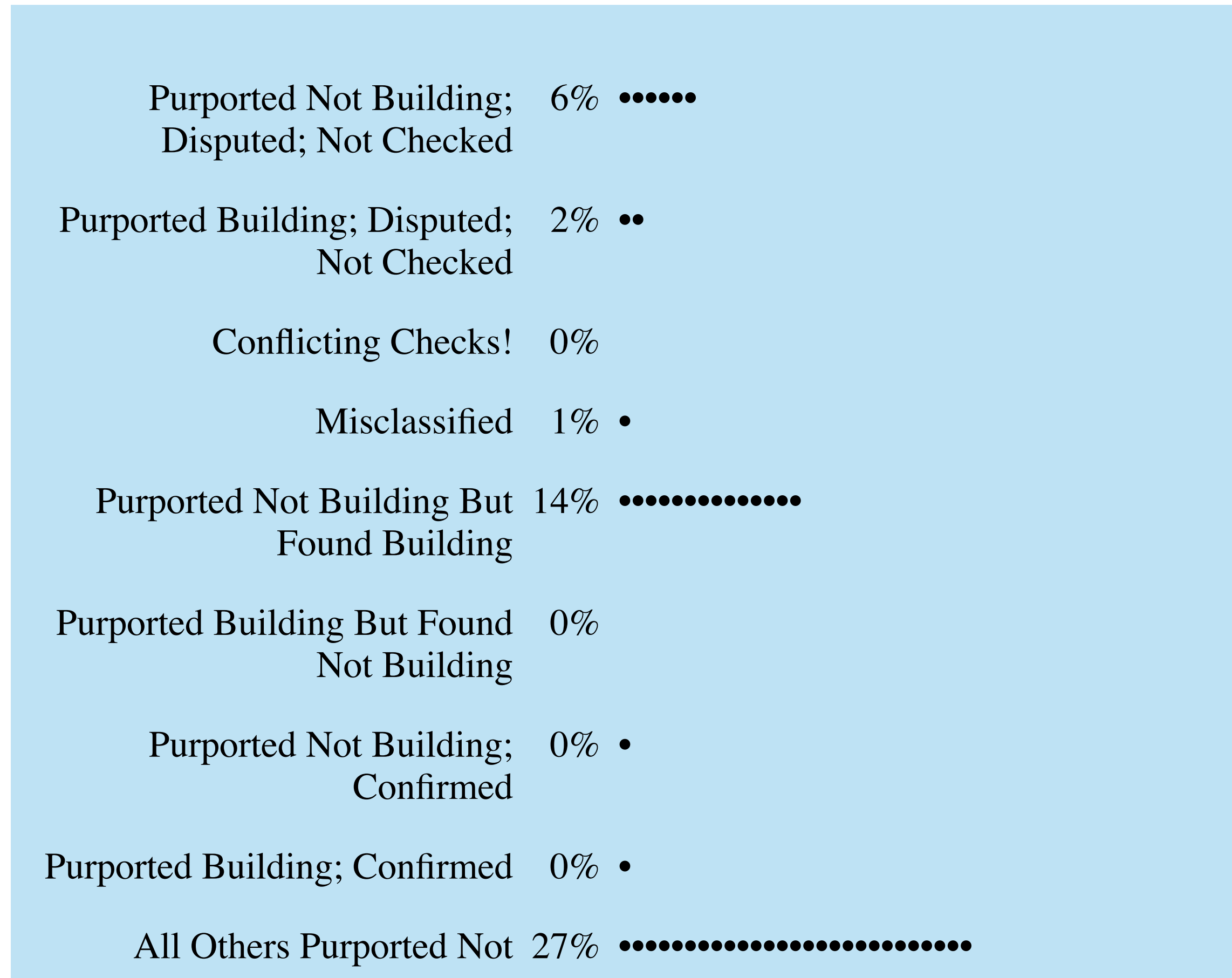
Excuses for not sharing

- Versioning
- Available Soon
- No Intention to Share
- Personnel Issues
- Lost Code
- Academic Tradeoffs
- Industrial Lab Tradeoffs
- Obsolete HW/SW
- Controlled Usage
- Privacy/Security
- Design Issues

[Collberg and Proebsting, 2015]

Examining 'Reproducibility in Computer Science'

- Repeat the experiment in reproducibility!
- Differences from original
- Shows issues with trying to classify experiments



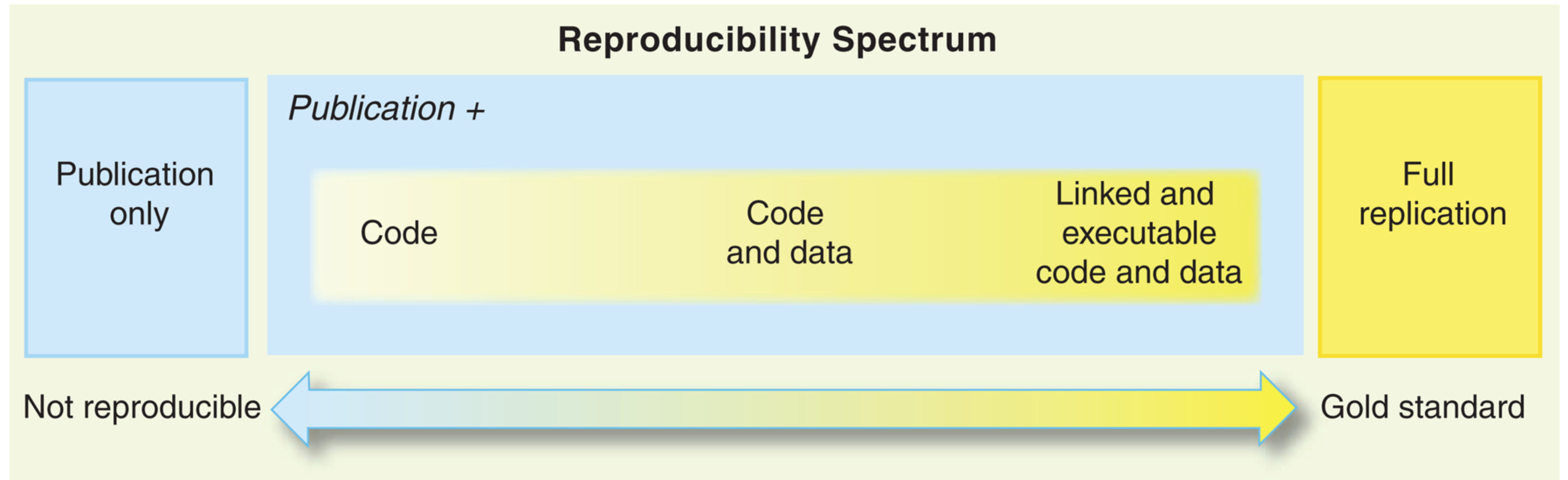
[S. Krishnamurthi et al.]

Reproducible Research

- Science is verified by replicating work independently
- Replication Issues:
 - Requires many resources to replicate (Sloan Digital Sky Survey)
 - Requires significant computing power (Climate Model Simulation)
 - Requires too much time or very specific circumstances (Environment Epidemiology)
- Reproducibility
 - Replication of the analysis based on the collected data (not replicating the data collection itself)
 - Better if we have the actual code or available executables

[R. D. Peng]

Reproducibility Spectrum



[R. D. Peng]

10 Rules for Reproducible Computational Research

- Rule 1: For Every Result, Keep Track of How It Was Produced
- Rule 2: Avoid Manual Data Manipulation Steps
- Rule 3: Archive the Exact Versions of All External Programs Used
- Rule 4: Version Control All Custom Scripts
- Rule 5: Record All Intermediate Results, When Possible in Standardized Formats

[Sandve et al., 2013]

10 Rules for Reproducible Computational Research

- Rule 6: For Analyses That Include Randomness, Note Underlying Random Seeds
- Rule 7: Always Store Raw Data behind Plots
- Rule 8: Generate Hierarchical Analysis Output, Allowing Layers of Increasing Detail to Be Inspected
- Rule 9: Connect Textual Statements to Underlying Results
- Rule 10: Provide Public Access to Scripts, Runs, and Results

[Sandve et al., 2013]

(Database) Reproducibility Research Topics

- Design and Management of Experiment Repositories
- Querying and Searching Experiments
- Mining Experiments

[J. Freire et al.]

Notebook Reproducibility

- Use notebooks from Github (~1 million)
 - Unambiguous cell order? 81.99%
- Study notebook dependencies
 - Dependencies Available? 13.72%
 - Dependencies Install? 5.03%
- Study notebook executability
 - Execute: 24.11% of unambiguous cell order
 - Matched results: 4.03%

[Pimentel et al., 2019]

Dataflow Notebooks: Resolve Notebook Ambiguities

```
In [d51f8eab]: import pandas as pd
df = pd.read_csv('guardian-top100-female-2019.csv')
```

df:

	Name	Rank	Position	Age on 1 Dec 2019	Nationality
0	Sam Kerr	1	Forward	26	Australia
...	...	...	...	...	...
99	Ludmila	100	Forward	25	Brazil

100 rows x 5 columns

```
In [full]: df = df.rename(columns={'Age on 1 Dec 2019': 'Age'})
```

df:

	Name	Rank	Position	Age	Nationality
0	Sam Kerr	1	Forward	26	Australia
...	...	...	...	...	...
99	Ludmila	100	Forward	25	Brazil

100 rows x 5 columns

```
In [over30]: df = df$full[df$full.Age >= 31]
```

df:

	Name	Rank	Position	Age	Nationality
2	Megan Rapinoe	3	Midfielder	34	USA
...	...	...	...	...	...
96	Cláudia Neto	97	Midfielder	31	Portugal

19 rows x 5 columns

```
In [under25]: df = df$full[df$full.Age <= 24]
```

df:

	Name	Rank	Position	Age	Nationality
3	Ada Hegerberg	4	Forward	24	Norway
...	...	...	...	...	...
98	Lena Oberdorf	99	Midfielder	17	Germany

25 rows x 5 columns

Assignment 5

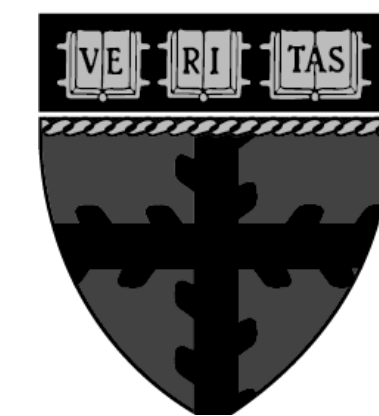
- Four parts
 - Loading Data
 - Spatial Analysis
 - Graph Analysis
 - Temporal Analysis
- Due at the end of the semester (April 22, 2021)
- Questions?

Final Exam

- Monday, April 26, 4:00-5:50pm, Online (Blackboard)
- Similar format
- More comprehensive (questions from topics covered in Test 1 & 2)
- Will also have questions from temporal data, provenance, reproducibility, machine learning
- Bring questions on Wednesday

Improving Databases

LEARNED AND SELF-DESIGNING DATA STRUCTURES

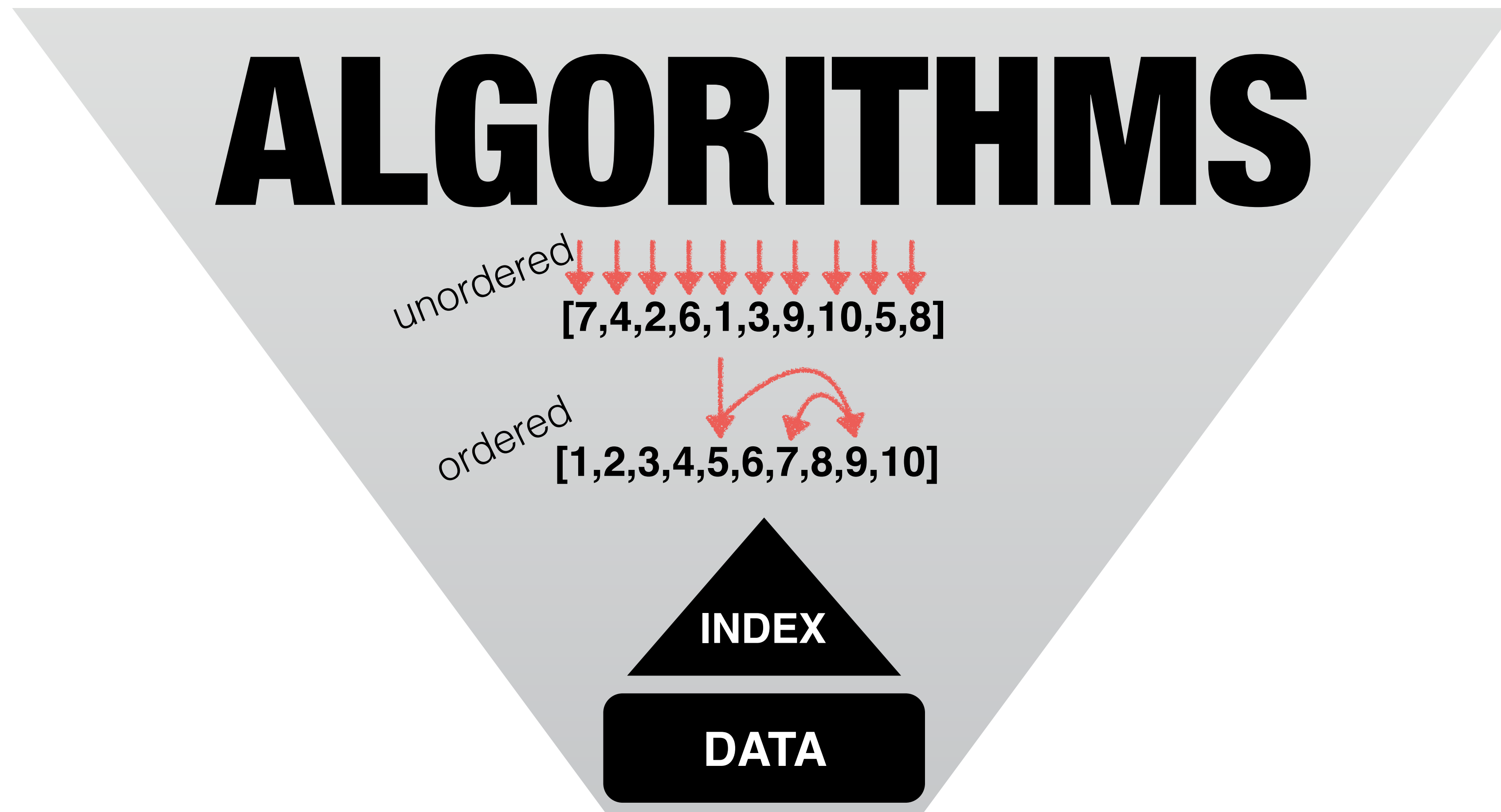


DASlab
@ Harvard SEAS

MIT DSAIL
Data Systems and AI Lab

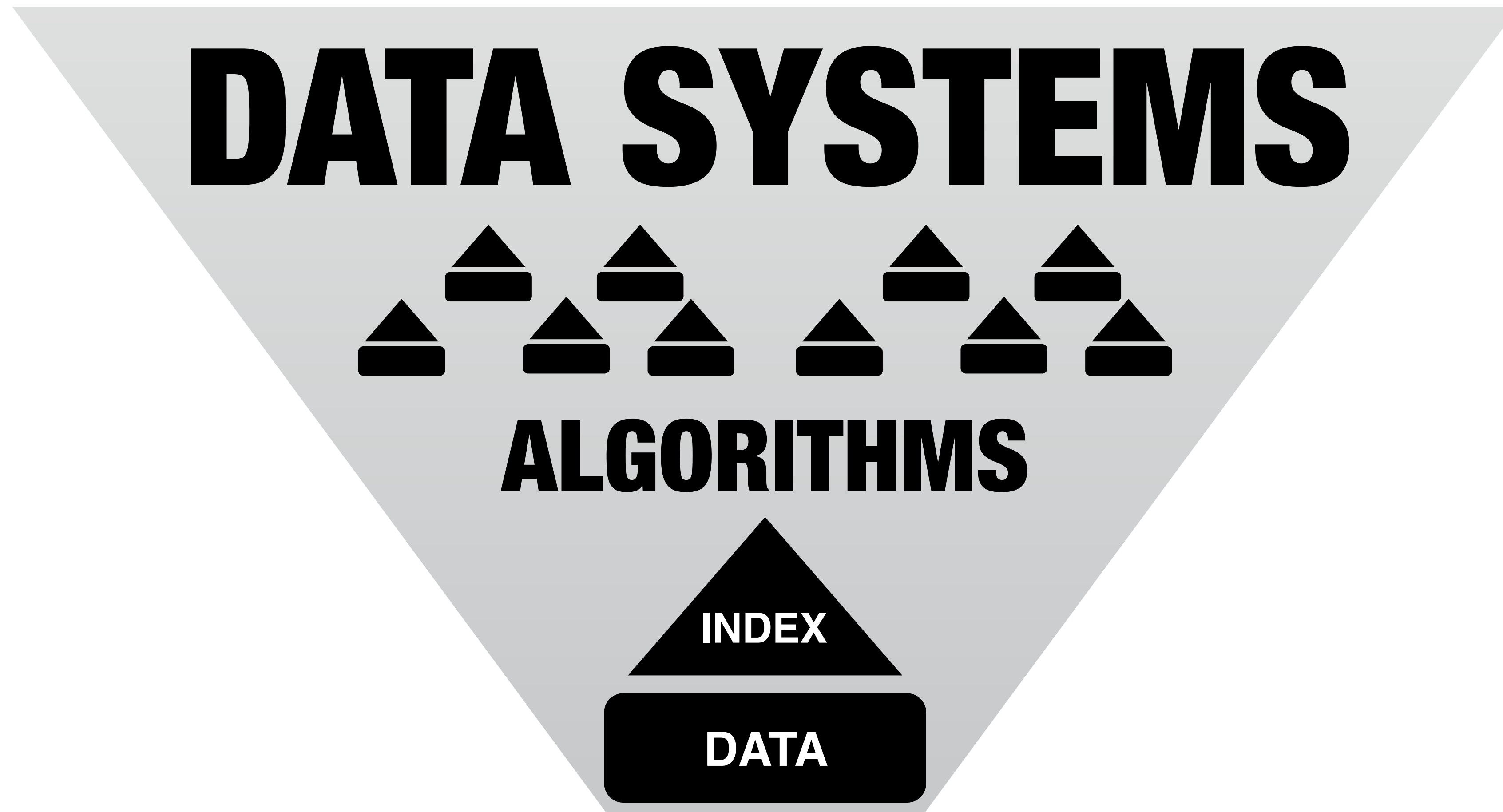
Stratos Idreos & Tim Kraska

Algorithms rely on the order of data



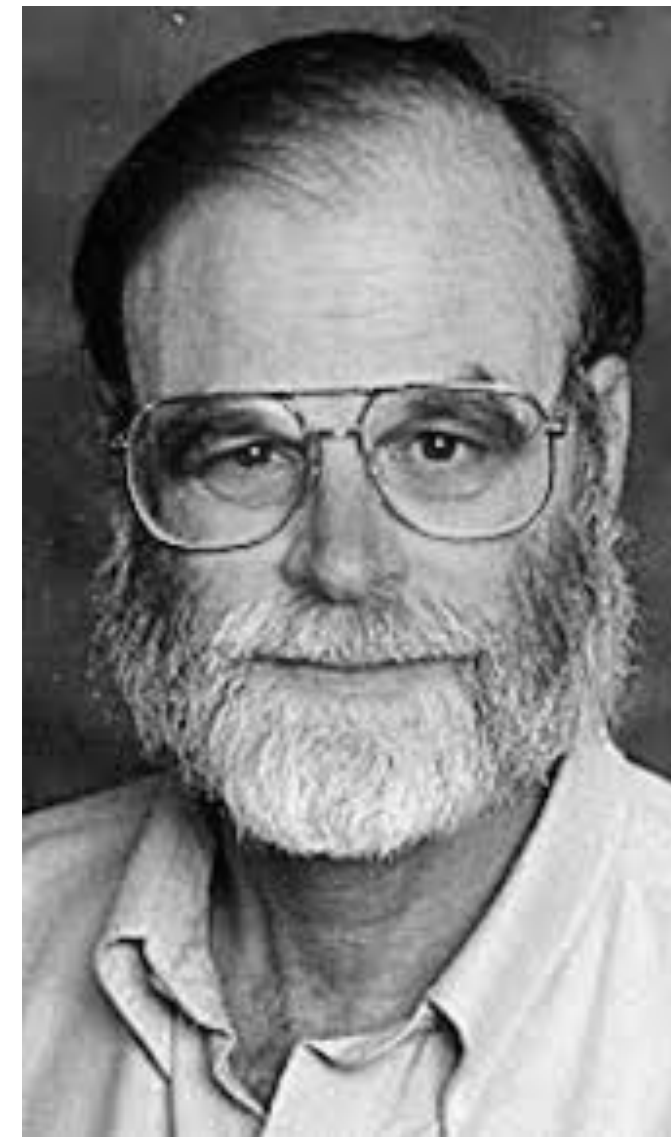
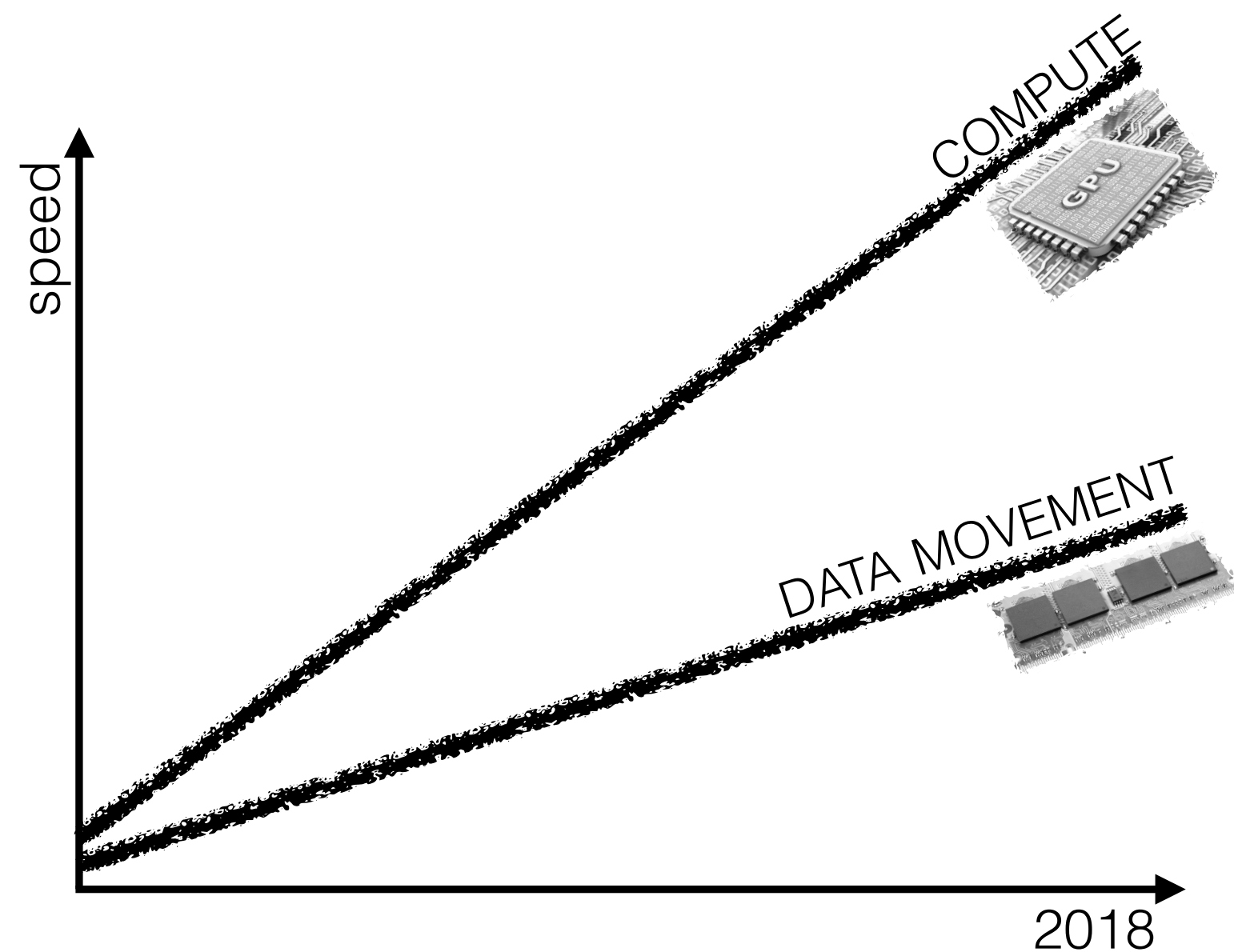
[S. Idreos, 2019]

Data systems rely on algorithms



[S. Idreos, 2019]

Data structures define performance



register = this room
caches = this city
memory = nearby city
disk = Pluto

Jim Gray, Turing Award 1998

[S. Idreos, 2019]

Database Questions

How do I make my **data system** run x times as fast?

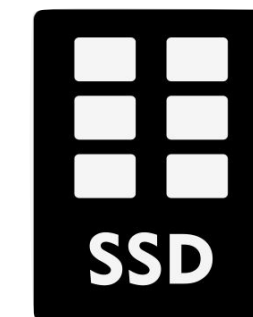


(sql,nosql,bigdata, ...)

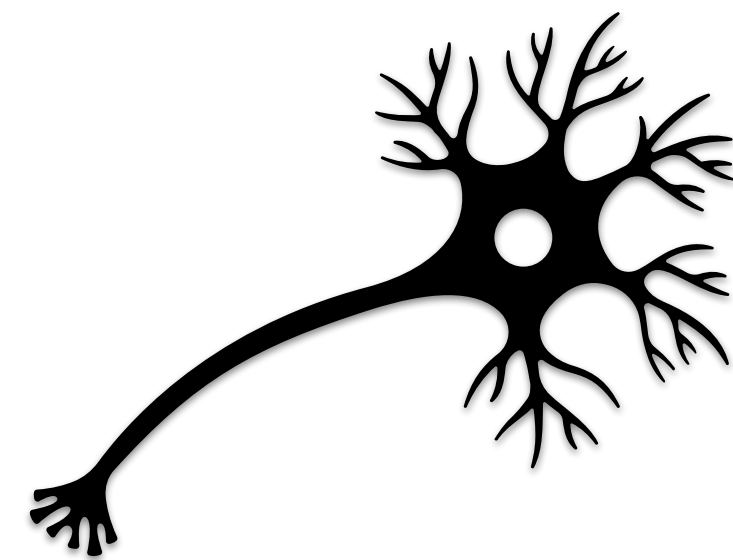


How do I minimize my **bill** in the **cloud**?

How do I extend the **lifetime** of my hardware?



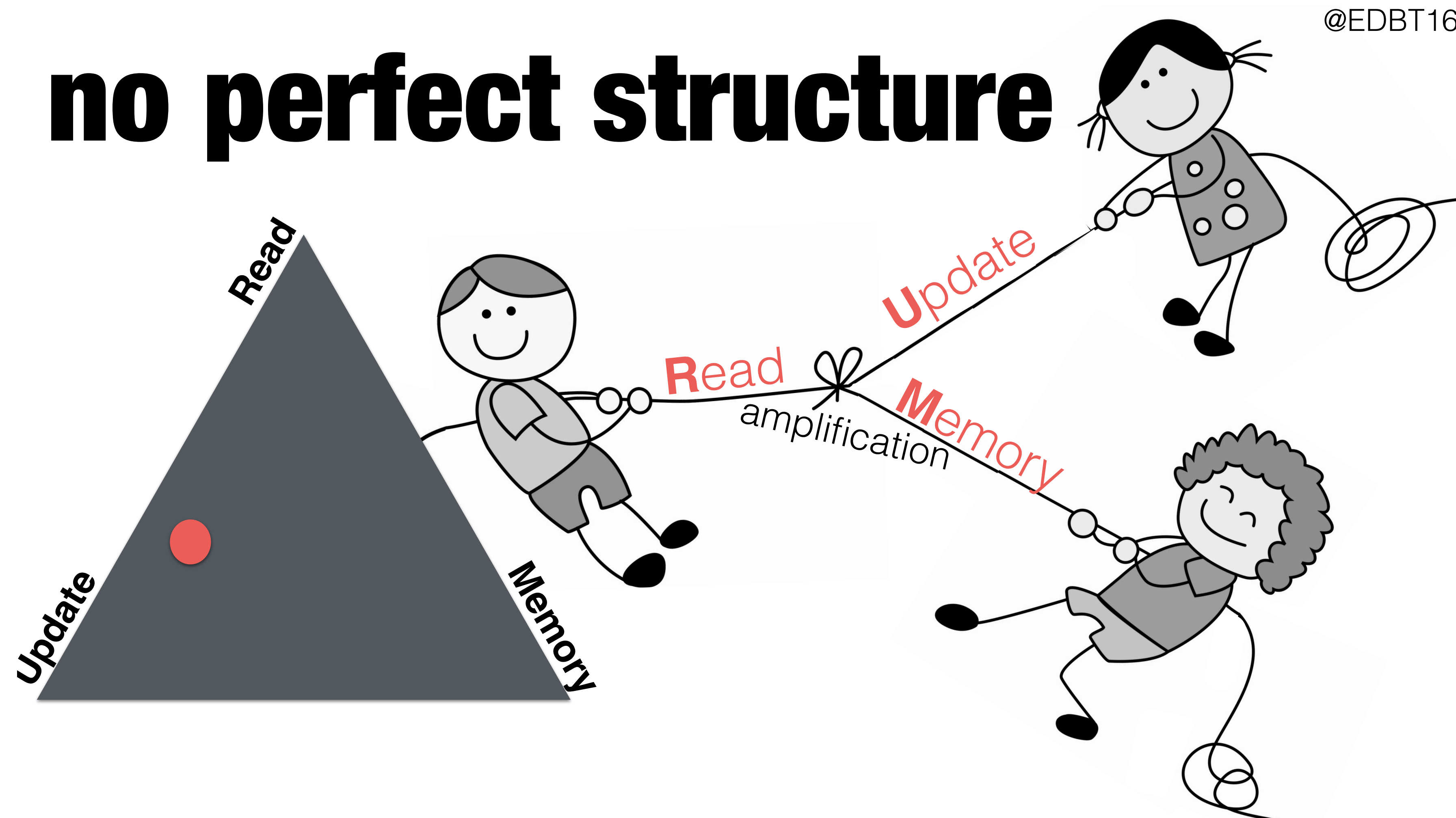
How to accelerate **statistics** computation for data science/ML?



How do I train my **neural network** x times faster?

[S. Idreos, 2019]

Tradeoffs in each structure

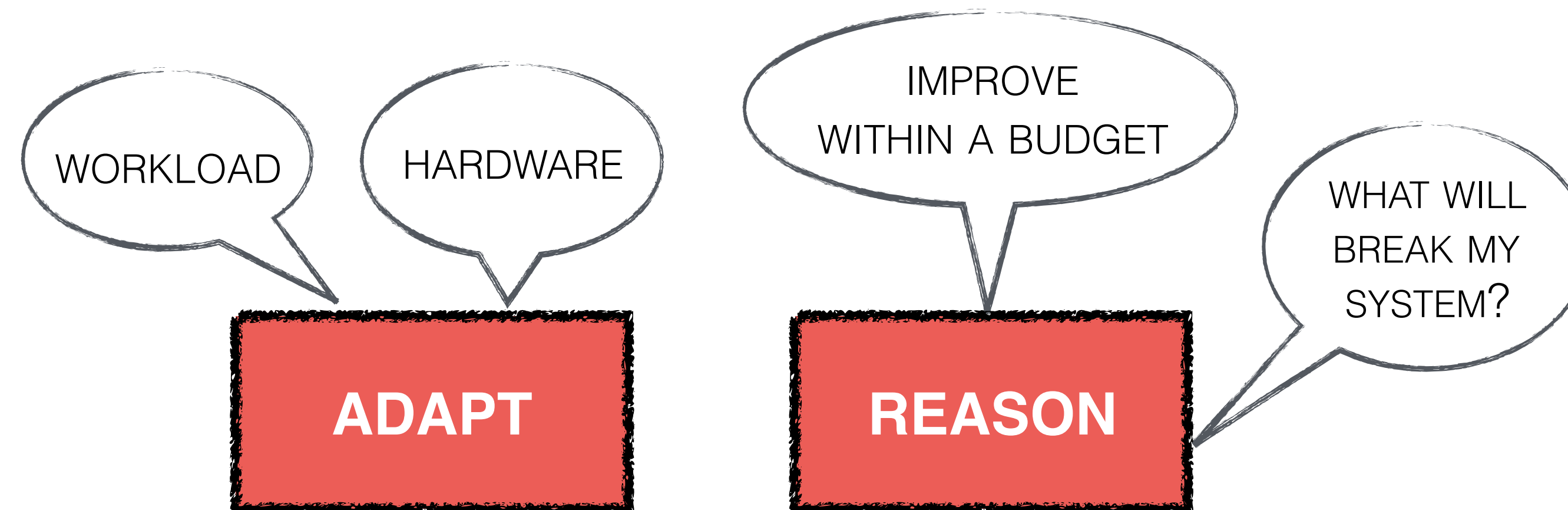


[S. Idreos, 2019]

New Applications Demand Change

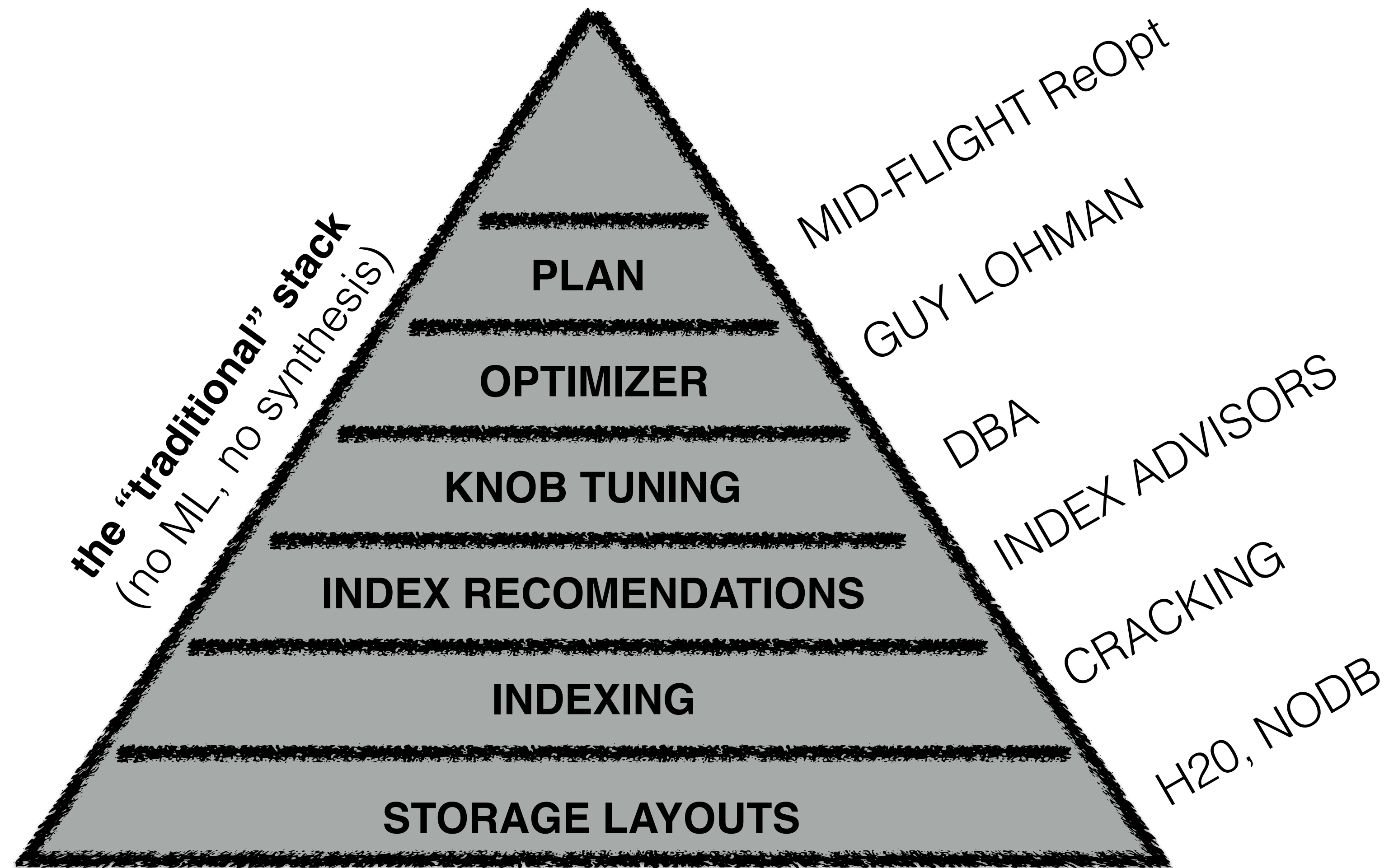
NEW APPLICATIONS 

existing systems need to change too



[S. Idreos, 2019]

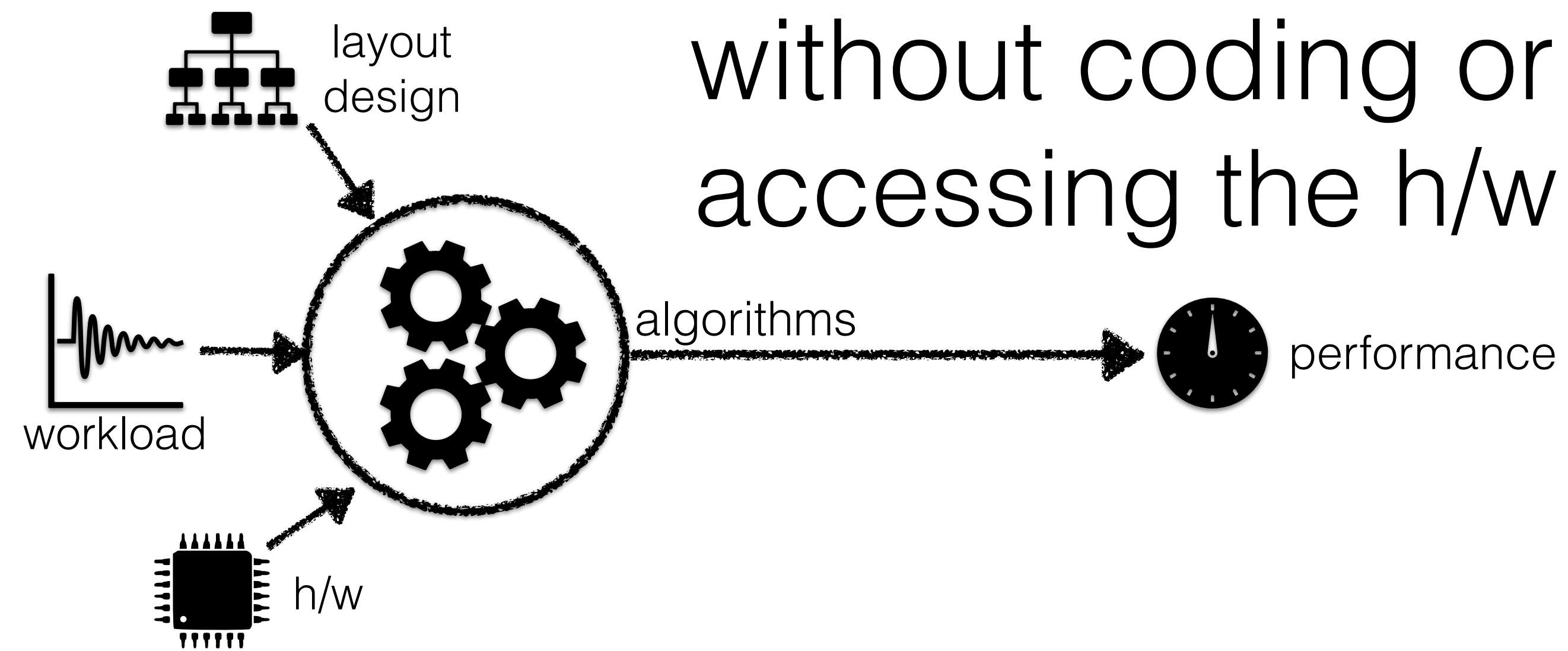
"Traditional" Database Research



[S. Idreos, 2019]

Self-designing systems

Data
Calculator



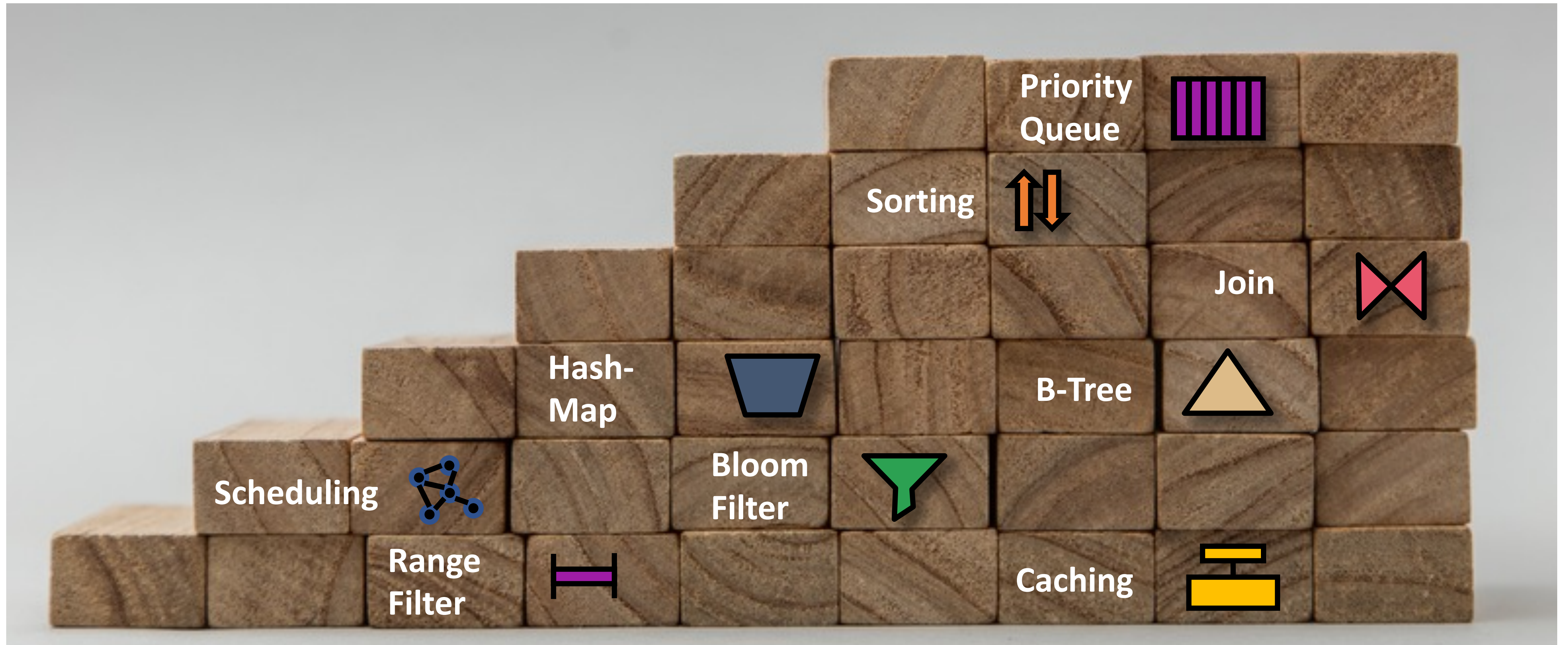
 **DASlab**
@ Harvard SEAS

[S. Idreos, 2019]

SageDB: a learned database system

T. Kraska, M. Alizadeh, A. Beutel, E. H. Chi, J. Ding, A. Kristo,
G. Leclerc, S. Madden, H. Mao, and V. Nathan

Learned Data Structures and Algorithms



Discussion

- Is this the future?
- What about comparison baselines?
- Lots of work being done in this area

Reminders

- Assignment 5 Due Thursday
- Final Exam Review Wednesday (come with questions!)
- Final Exam on Monday, April 26 from 4-5:50pm (Online)