Advanced Data Management (CSCI 640/490)

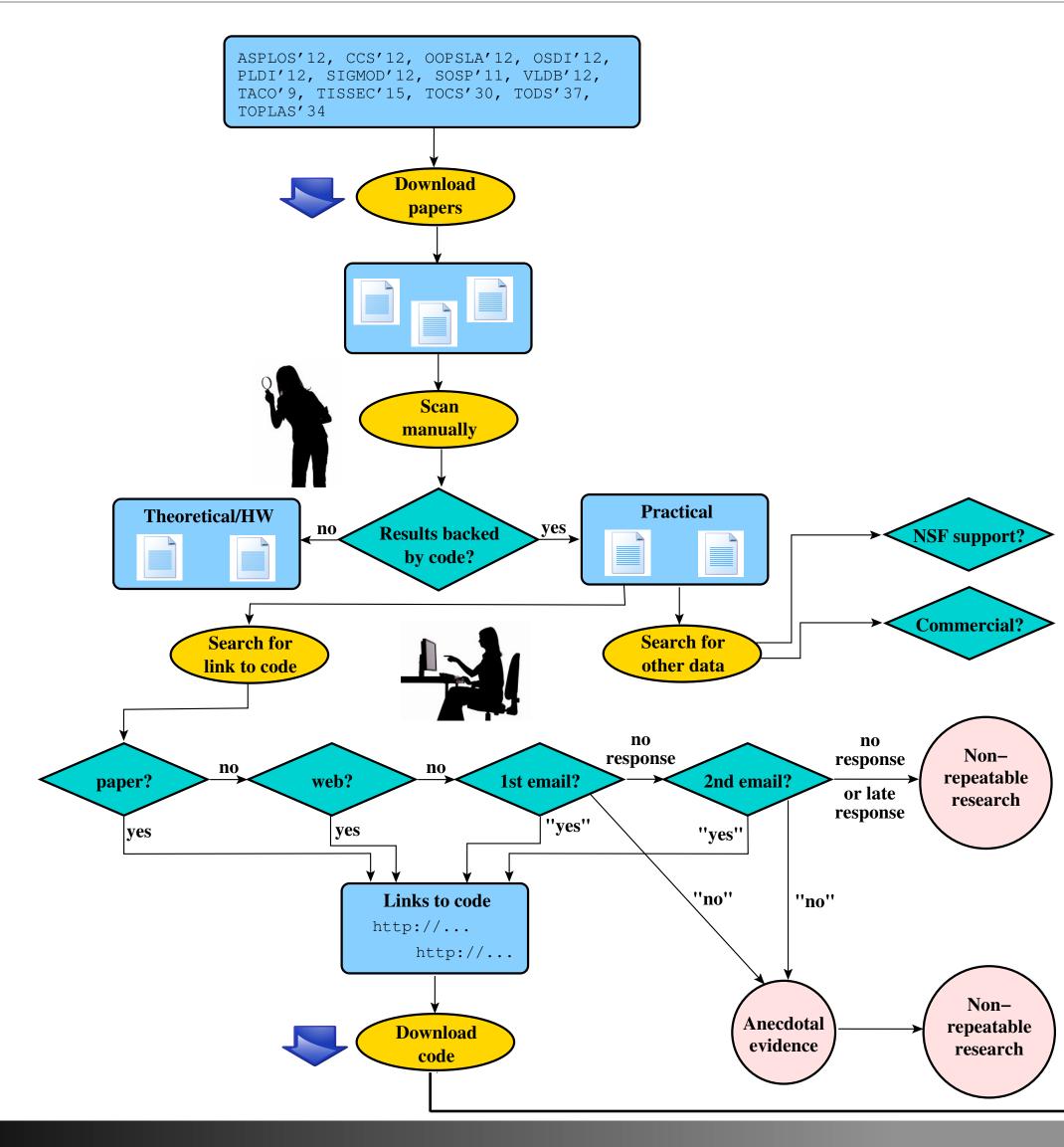
Machine Learning in Databases

Dr. David Koop

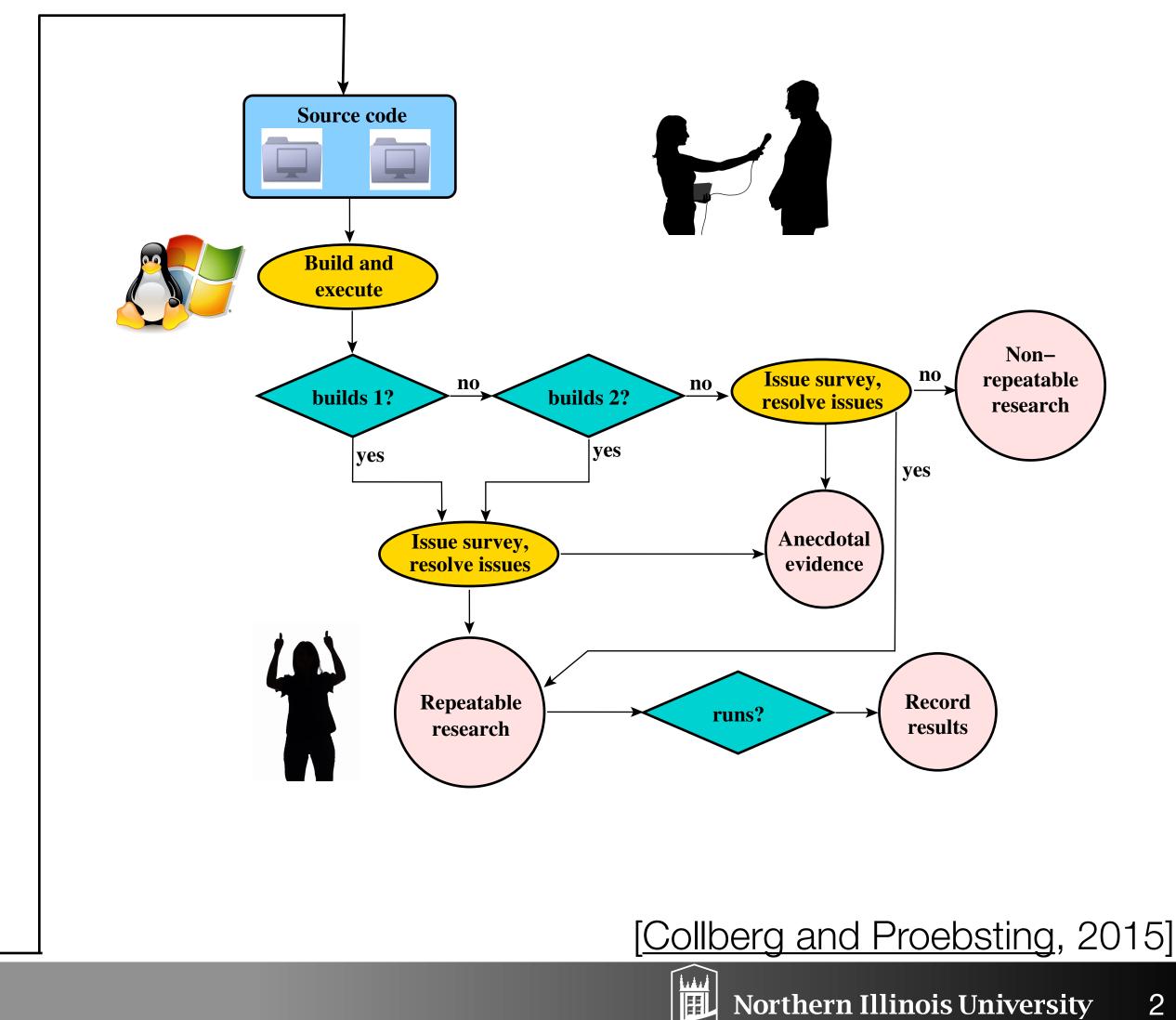




Checking Computational Results in Systems



D. Koop, CSCI 640/490, Spring 2023



Northern Illinois University

NIU





Repeatability Results

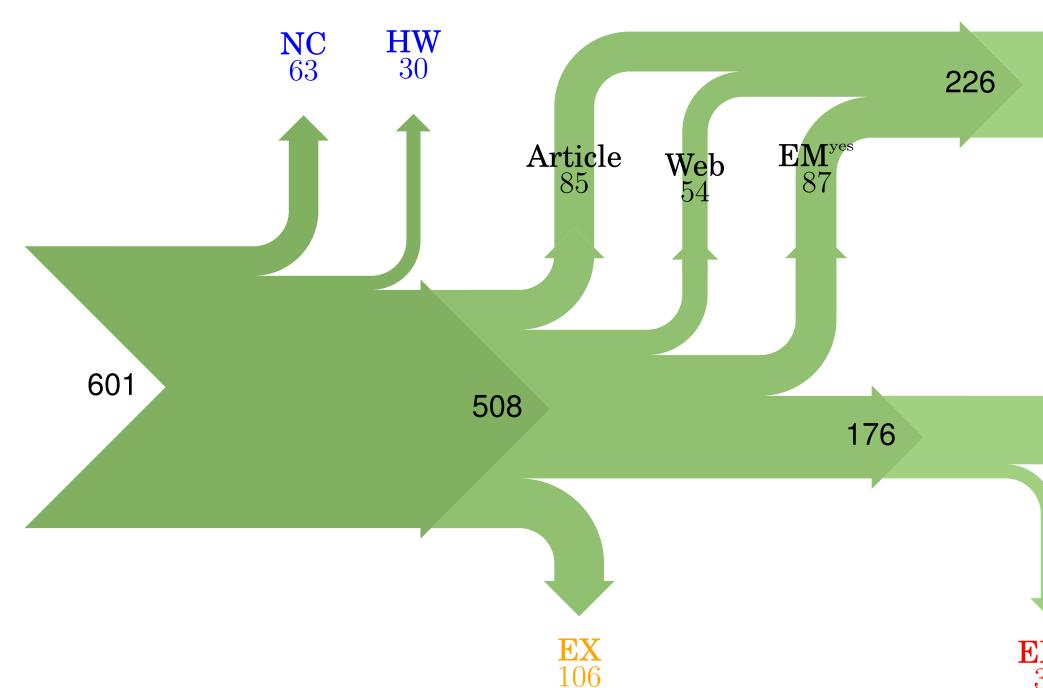
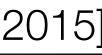


Figure 11: Study result. Blue numbers represent papers that were excluded from consideration, green numbers papers that are weakly repeatable, red numbers papers that are non-weakly repeatable, and orange numbers represent papers that were excluded (due to our restriction of sending at most one email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

OK ^{≤30} OK 130 64	> ³⁰ OK ^{Auth} 23	
	Notation	Number of papers
Build fails 9	HW	excluded due to replication requiring special hardware
	NC	excluded due to results not being backed by code
	EX	excluded due to overlapping author lists
	BC	where the results are backed by code
	Article	where code was found in the paper itself
	Web	where code was found through a Web search
	EM yes	where the author provides code after receiving an email message
	EM ^{no}	where the author responds to an email message saying code cannot be provided
	EMø	where the author does not respond to email requests within two months
	OK ^{≤30}	where code is available and we succeed in building the system in \leq 30 minutes
	OK >30	where code is available and we succeed in building the system in >30 minutes
	OK ^{Auth}	where code is available and we fail to build, and the author says the code builds with reasonable effort
$\begin{array}{ccc} \mathbf{M}^{\emptyset} & \mathbf{E}\mathbf{M}^{\mathrm{no}} \\ 30 & 146 \end{array}$	Fails	where code is available and we fail to build, and the author says the code may have problems building













Excuses for not sharing

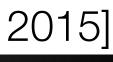
- Versioning
- Available Soon
- No Intention to Share
- Personnel Issues
- Lost Code
- Academic Tradeoffs
- Industrial Lab Tradeoffs
- Obsolete HW/SW
- Controlled Usage
- Privacy/Security
- Design Issues

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Examining 'Reproducibility in Computer Science'

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

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	Γ)i	

All Others Purported Not 27%

- ported Not Building; 6% ••••• sputed; Not Checked
- Purported Building; Disputed; 2% •• Not Checked
 - Conflicting Checks! 0%
 - Misclassified 1% •
 - Purported Not Building But 14% ••••••••• Found Building
- Purported Building But Found 0% Not Building
 - Purported Not Building; 0% Confirmed
- Purported Building; Confirmed 0% •









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 _

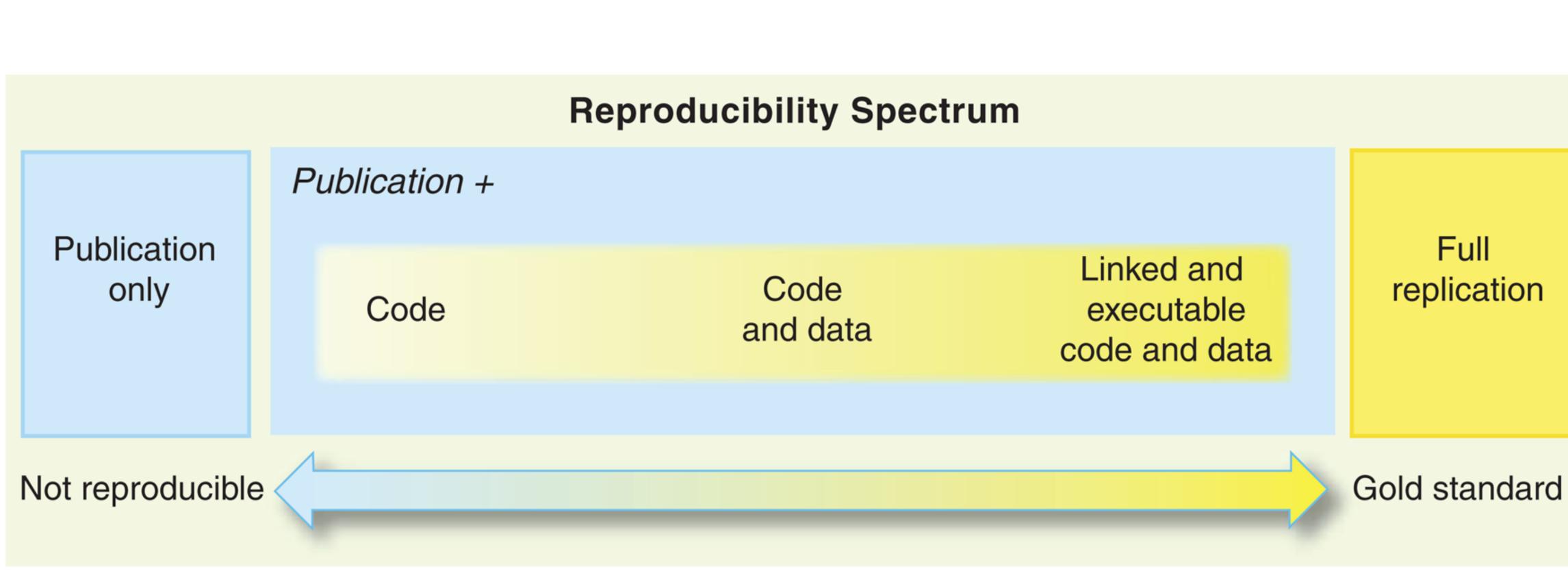








Reproducibility Spectrum



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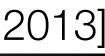
7

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





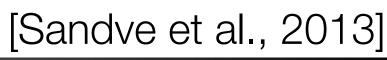


10 Rules for Reproducible Computational Research

- 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









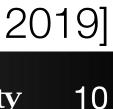


Notebook Reproducibility

- 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%







Dataflow Notebooks: Resolve Notebook Ambiguities

In [d51f8eab]:	<pre>import pandas as pd df = pd.read_csv('guardian-top100-female-2019.csv')</pre>	In [over30]:	<pre>df = df\$full[df\$full.Age >= 31]</pre>
df:	Name Rank Position Age on 1 Dec 2019 Nationality	df:	Name Rank Position Age Nationality
	0 Sam Kerr 1 Forward 26 Australia		2 Megan Rapinoe 3 Midfielder 34 USA
	···· ··· ··· ···		••• •• ••• •••
	99 Ludmila 100 Forward 25 Brazil		96 Cláudia Neto 97 Midfielder 31 Portugal
	100 rows × 5 columns		19 rows × 5 columns
In [full]:	<pre>df = df.rename(columns={'Age on 1 Dec 2019': 'Age'})</pre>	In [under25]: df:	<pre>df = df\$full[df\$full.Age <= 24]</pre>
df:	Name Rank Position Age Nationality		Name Rank Position Age Nationality
ur.	0 Sam Kerr 1 Forward 26 Australia		3 Ada Hegerberg 4 Forward 24 Norway
			···· ··· ··· ··· ···
	99 Ludmila 100 Forward 25 Brazil		98 Lena Oberdorf 99 Midfielder 17 Germany
	100 rows × 5 columns		25 rows × 5 columns

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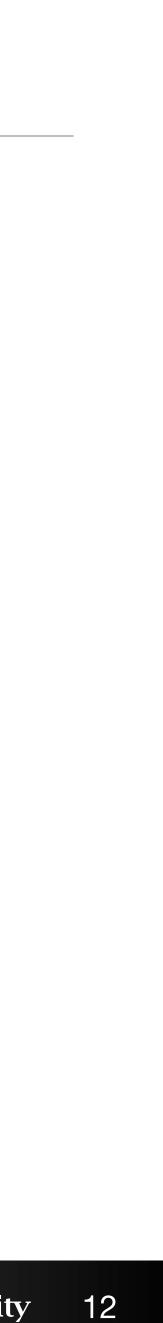


/ **11**

<u>Assignment 5</u>

- Chicago Bike Sharing Data
 - Spatial Analysis
 - Temporal Analysis
 - Graph Database (neo4j)





Final Exam

- Wednesday, May 10, **8:00**-9:50pm, PM 253
- Similar format
- More comprehensive (questions from topics covered in Test 1 & 2)
- Will also have questions from graph/spatial/temporal data, provenance, reproducibility, machine learning



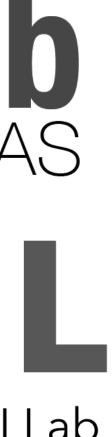


Improving Databases





LEARNED AND **SELF-DESIGNING** DATA STRUCTURES Data Systems and AI Lab Stratos Idreos & Tim Kraska



Algorithms rely on the order of data





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ALGORITHMS

[7,4,2,6,1,3,9,10,5,8]











Data systems rely on algorithms

DATA SYSTEMS ALGORITHMS

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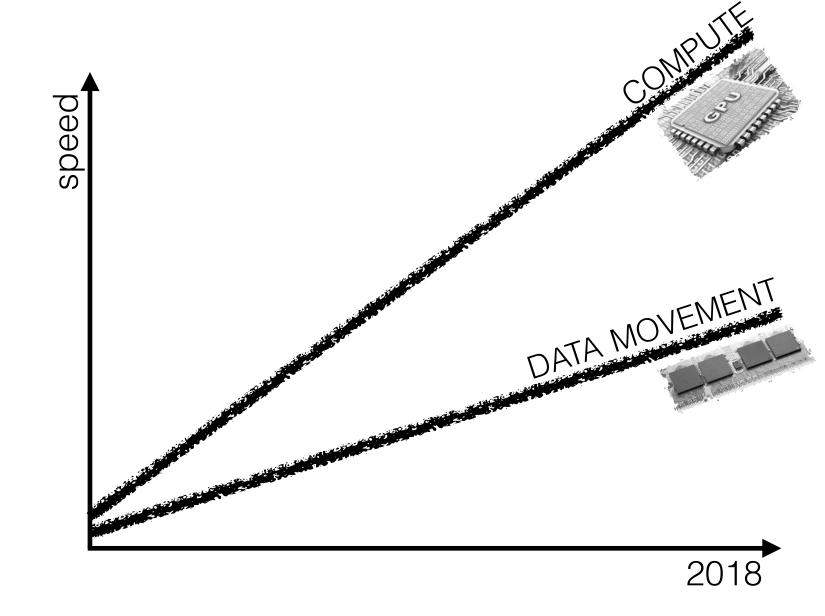




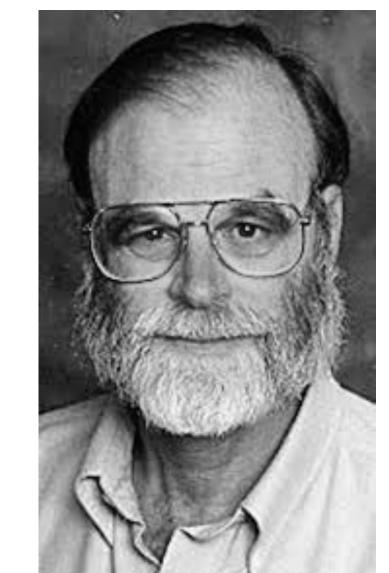


17

Data structures define performance



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

Jim Gray, Turing Award 1998







18

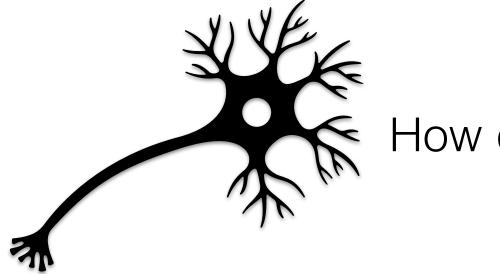
Database Questions

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



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

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



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

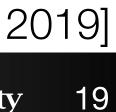




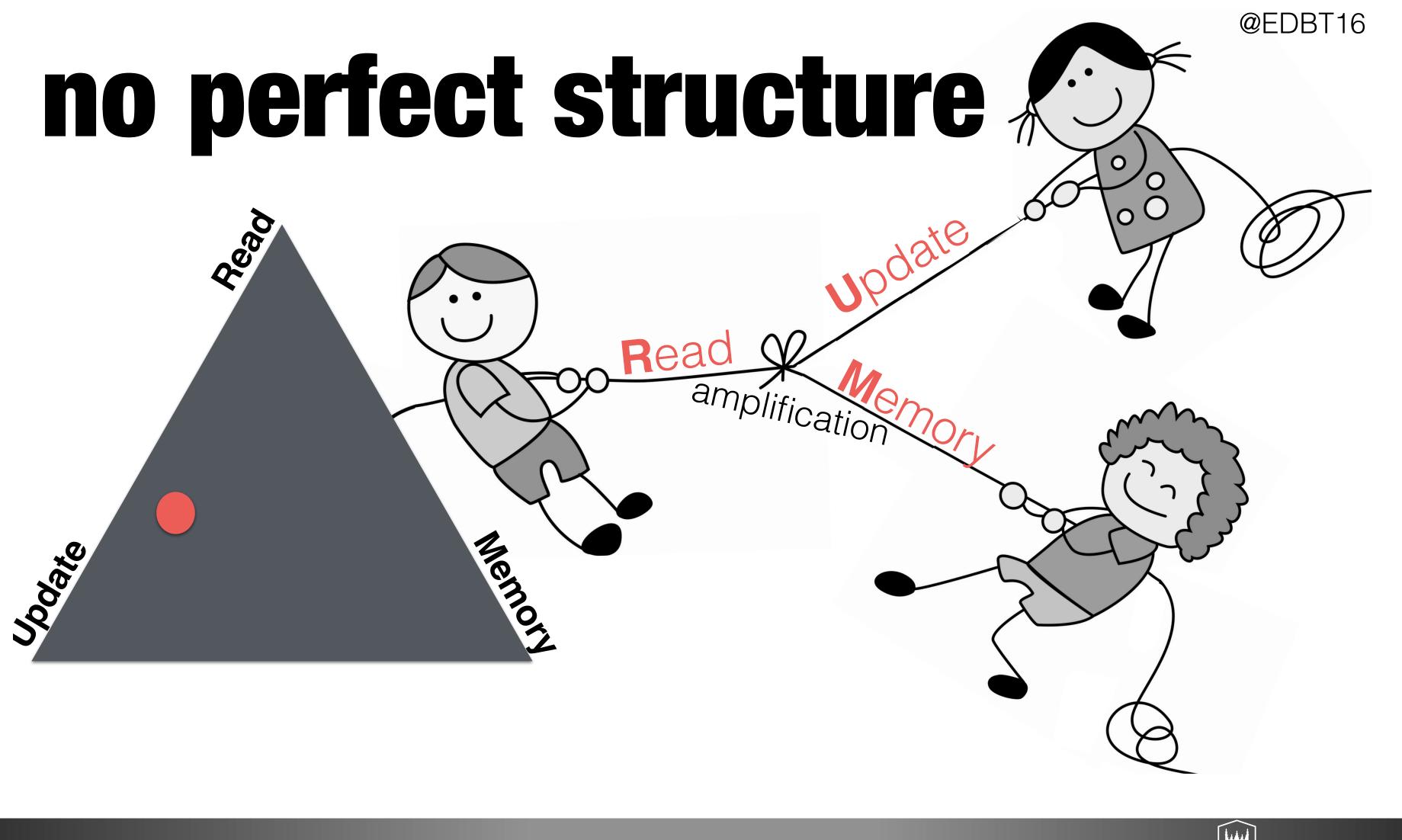
How do I train my **neural network** x times faster?







Tradeoffs in each structure







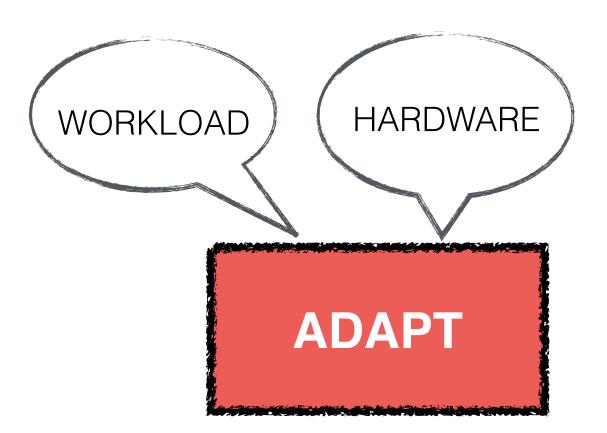






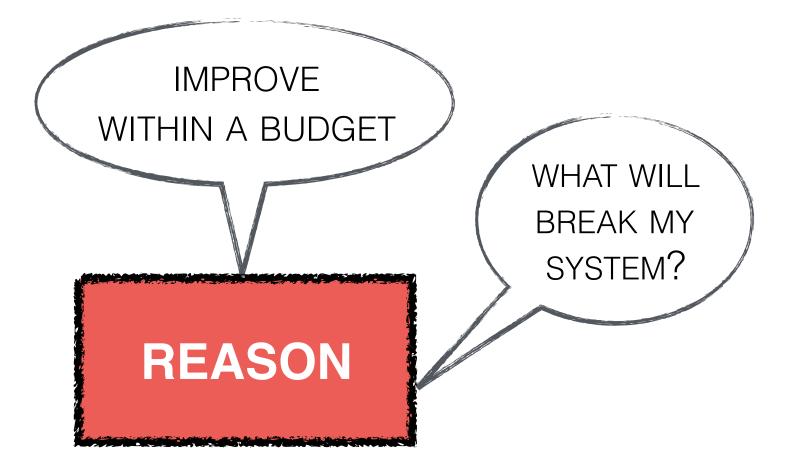
New Applications Demand Change

existing systems need to change too









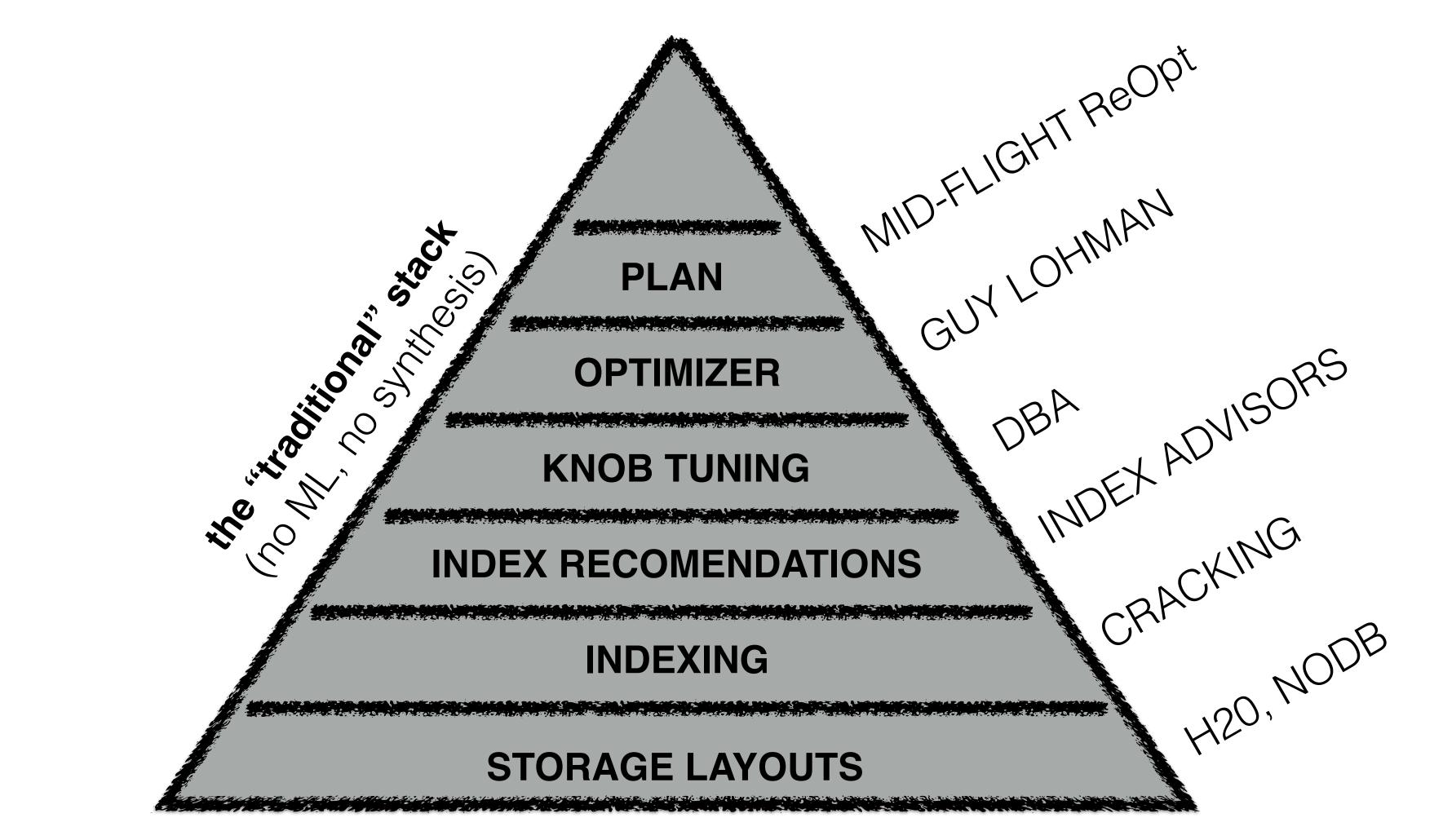








"Traditional" Database Research





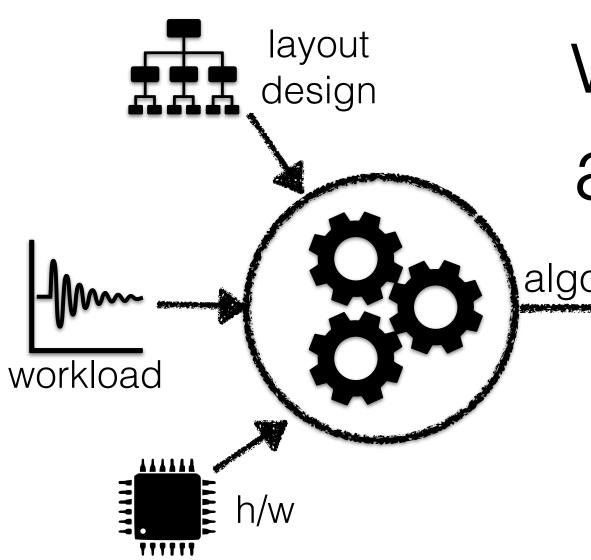






Self-designing systems







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without coding or accessing the h/w

algorithms



performance











SageDB: a learned database system

G. Leclerc, S. Madden, H. Mao, and V. Nathan

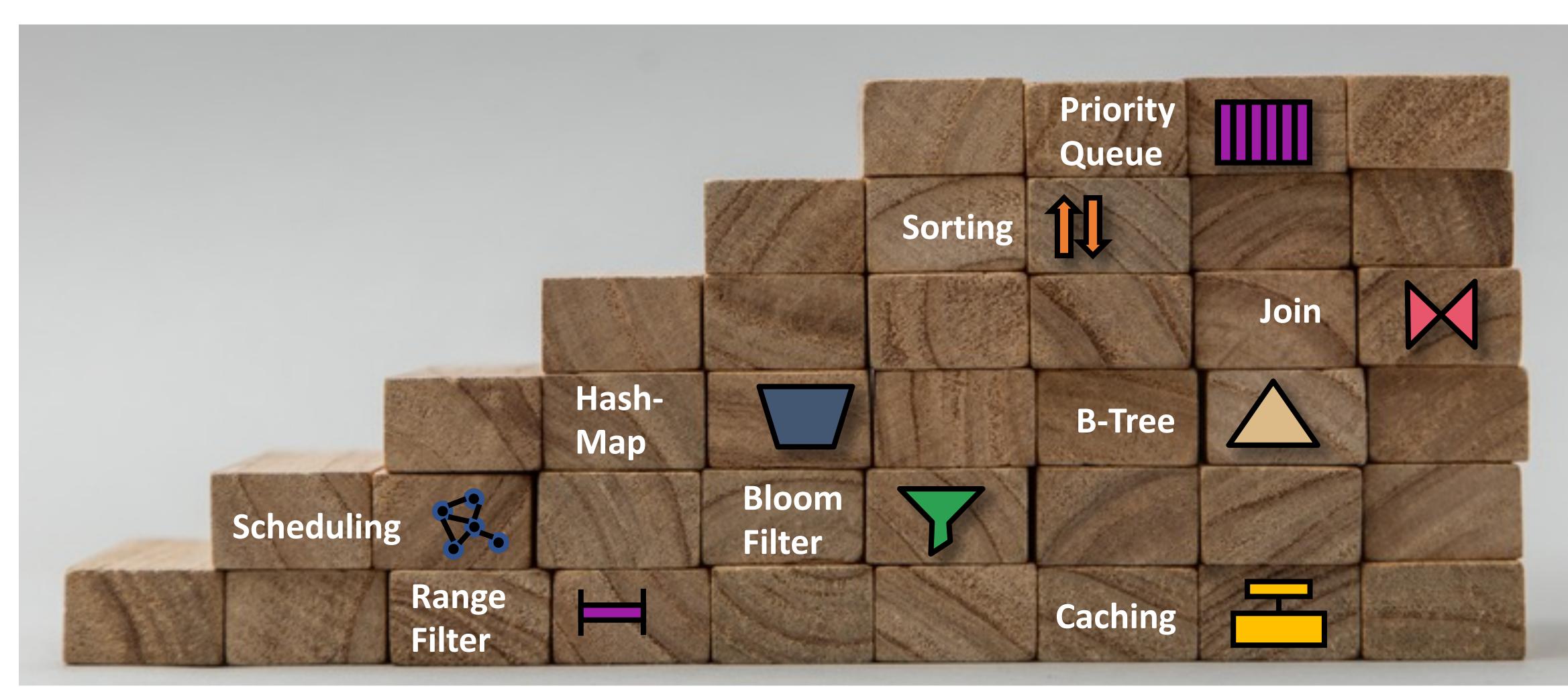
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T. Kraska, M. Alizadeh, A. Beutel, E. H. Chi, J. Ding, A. Kristo,

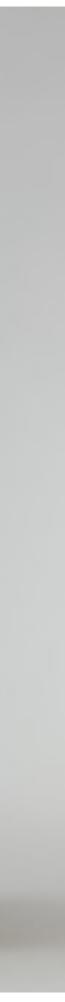




Learned Data Structures and Algorithms









Discussion

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



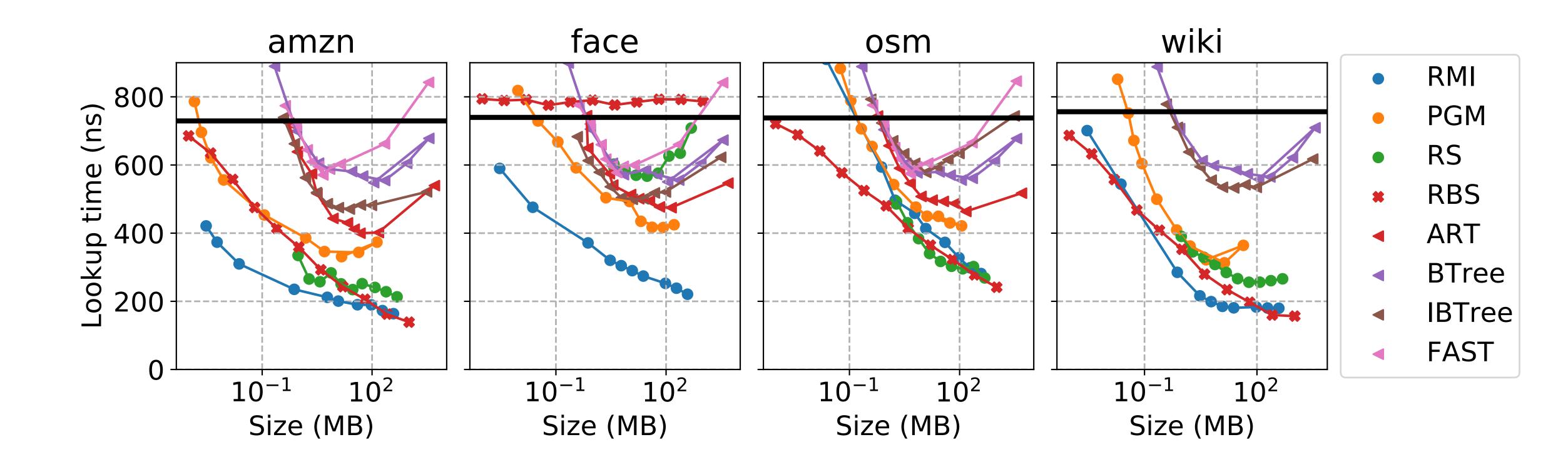








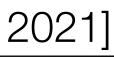
Benchmarking Learned Indexes



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[<u>R. Marcus et al.</u>, 2021]

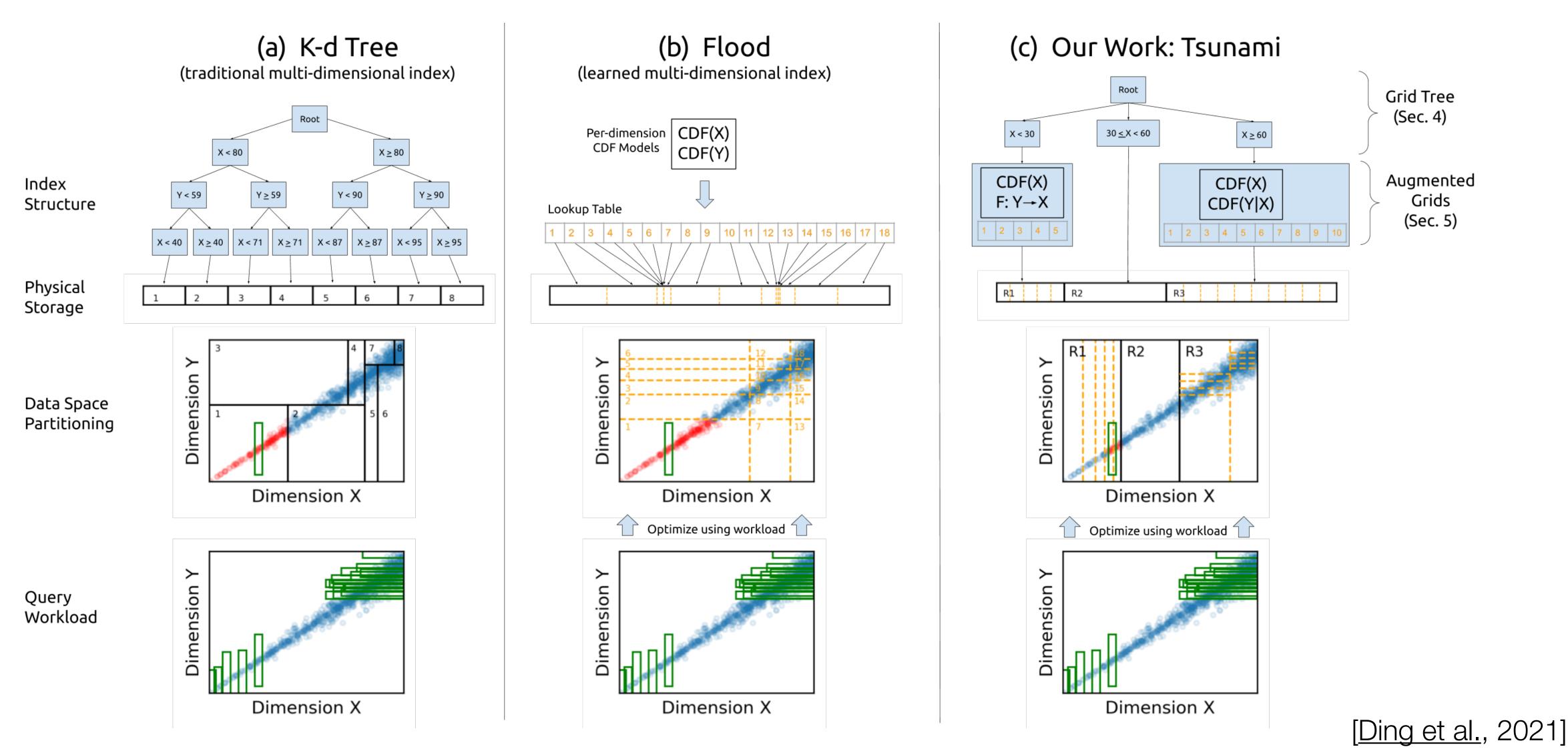








Multi-Dimensional Indexing



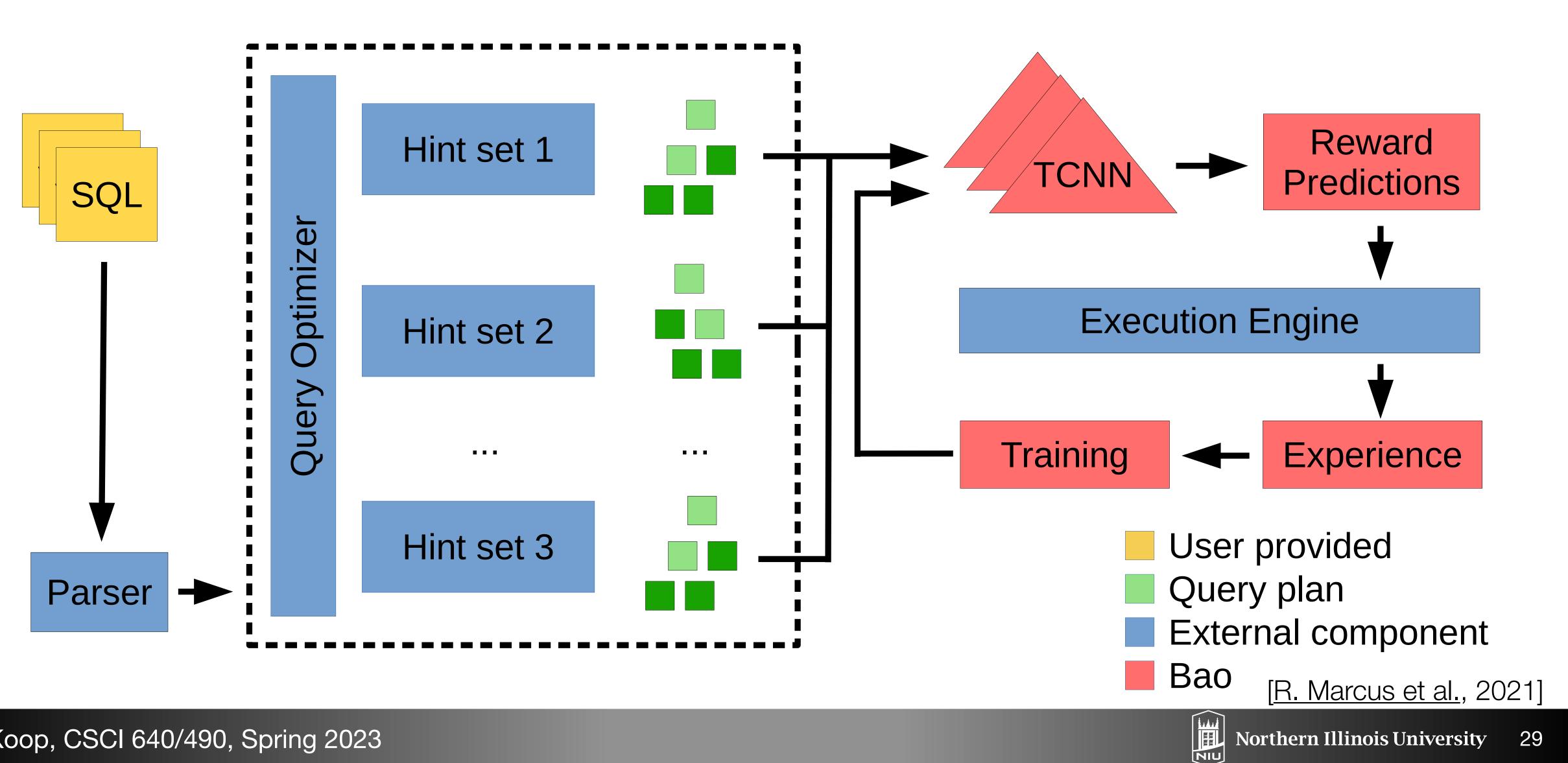








Query Optimization



Reminders

- Final Exam Review Wednesday (come with questions!)
- Final Exam on Wednesday, May 10 from 8:00-9:50am





