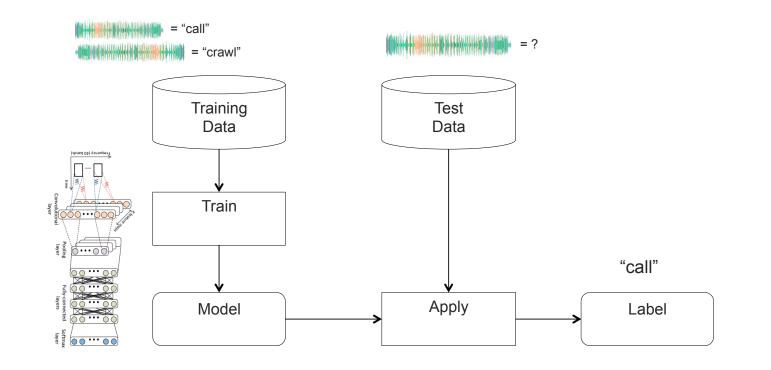
ONLINE MACHINE LEARNING AND DATA MINING

EDO LIBERTY

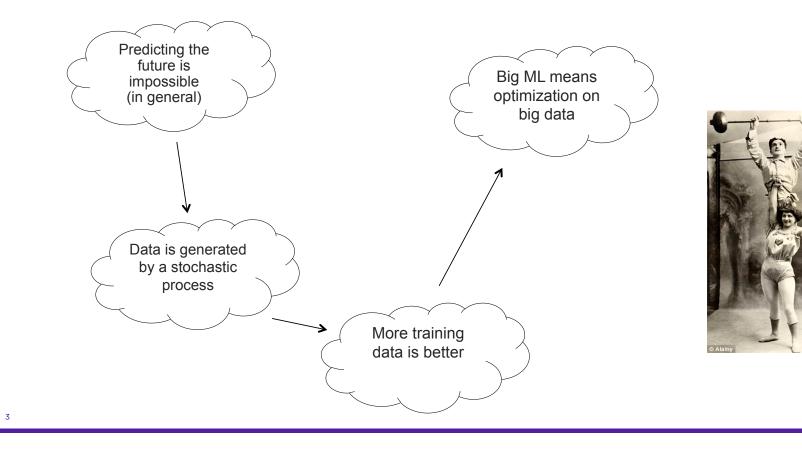


STANDARD MACHINE LEARNING SETTING



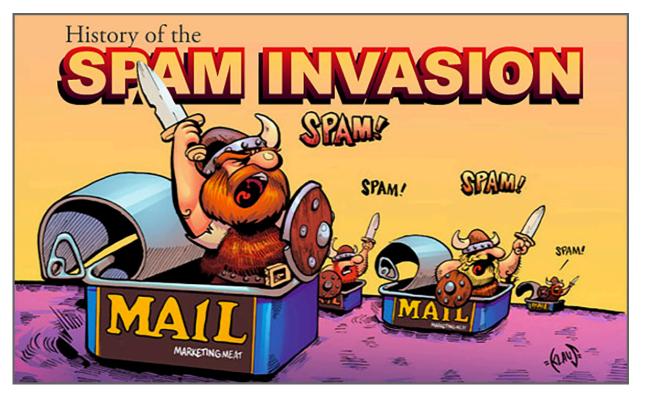


STANDARD MACHINE LEARNING SETTING



YAHOO!

MORE DATA IS OFTEN WORSE (MORE DATA = OLDER DATA)



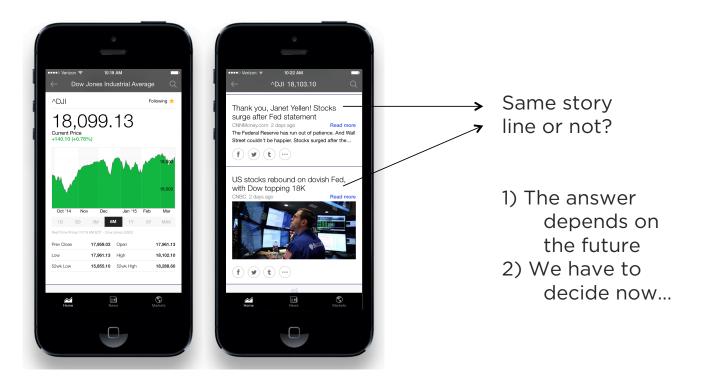


OUR ACTIONS HEAVILY INFLUENCE THE DATA

YAHOO!	kitchen tables with benches				Search		
	Web Images	Video	Anytime ~				
	Also try: small kitchen tables with benches						
	Ads related to: kitchen tables with benches						A FR
	Deals on Dining Tables - Shop Dining Tables Online. www.raymourflanigan.com/Dining-Room Shop Dining Tables Online. Enjoy Store-Wide Savings Today! 490 Fulton Street, New York, NY (347) 416-5019 Directions Financing Options - Store Locator - Dining Chairs - Glass Tables Kitchen Benches on Sale - 20%-50% Off Kitchen Benches. www.ATGStores.com/DiningBenches 20%-50% Off Kitchen Benches. 7 Day Customer Service & Free Shipping! Brands: Homelegance, Liberty Furniture, Nuevo Living, Sunny Designs and more					Ad	
							erson 62 g Table,.
						\$799 west e	
	Dining Benche	s With Arms	Traditio	onal Dining Bench			
	Dining Benche	s on Sale	Moderr	n Dining Benches		STATES	Manhi
	Kitchen Islands	s & Carts	Backles	ss Dining Benches	5	X	

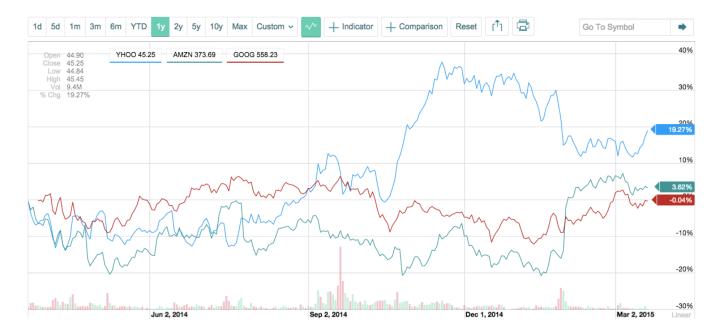


THE FUTURE IS OFTEN NOT LIKE THE PAST!





HAVING "A MODEL" IS COMPLETELY UNIMPORTANT



Elements of information theory, Cover, 1991 Efficient algorithms for universal portfolios, Kalai, Vempala, 2003 Efficient Algorithms for Online Game Playing and Universal Portfolio Management, Agarwal, Hazan, 2006



ONLINE ALGORITHMS (DECISION MAKING WITHOUT PREDICTING)







Rent: x\$ /day

9

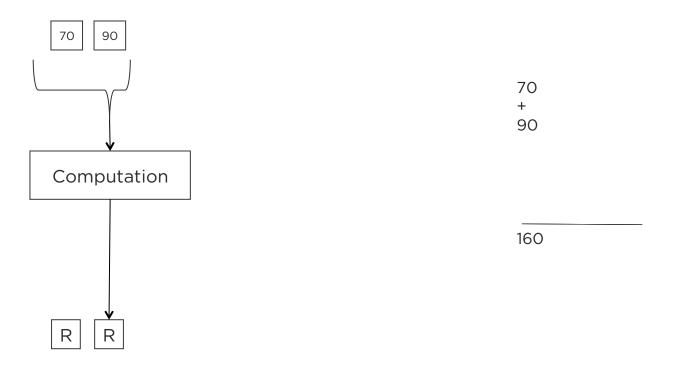


Buy: 1000\$

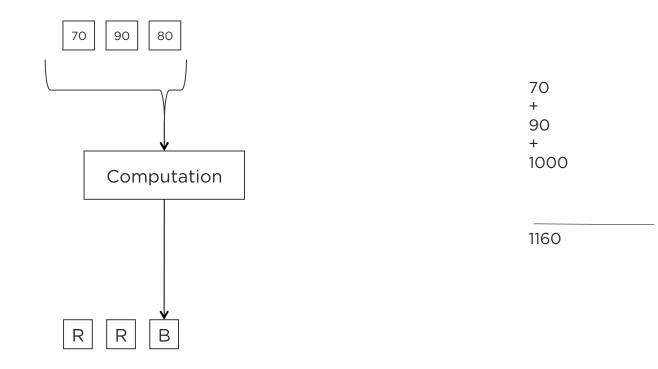




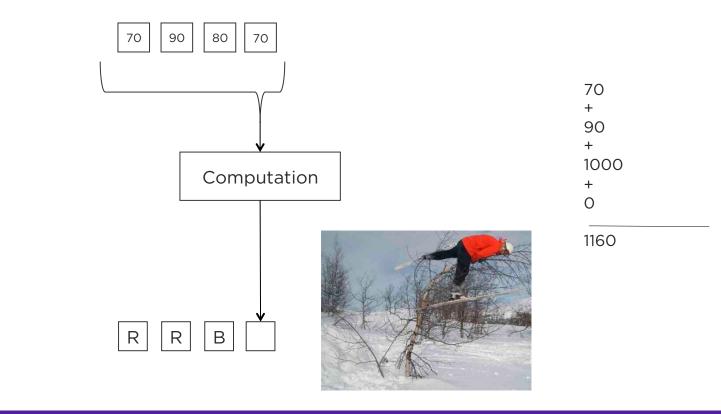




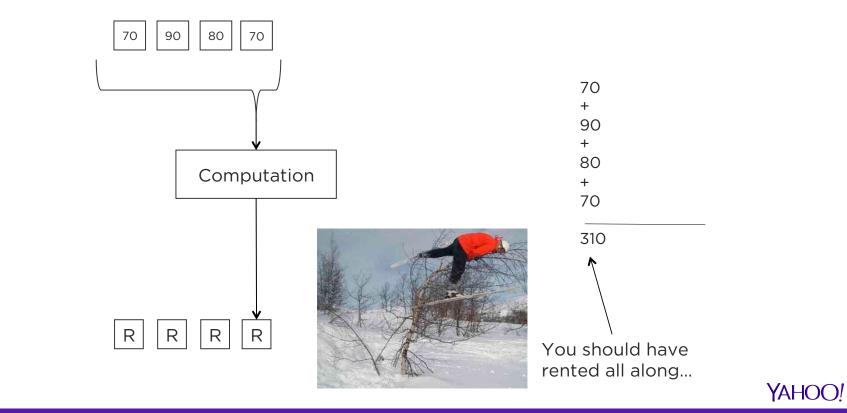


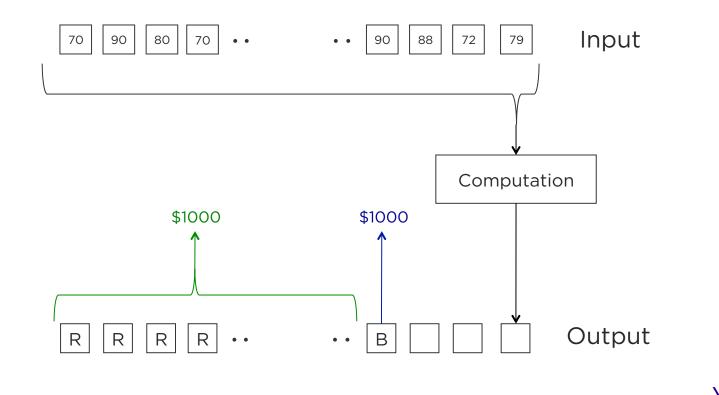




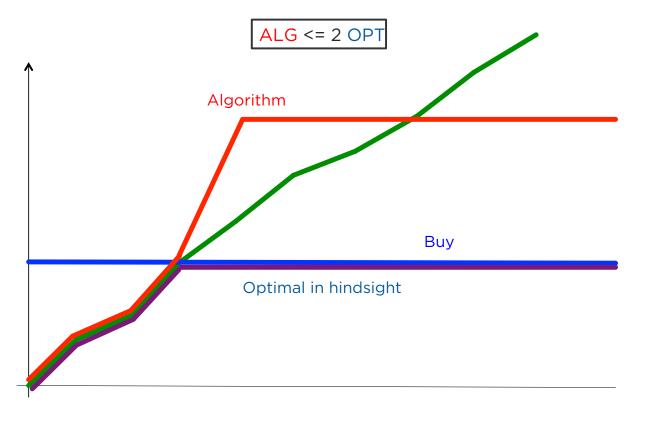










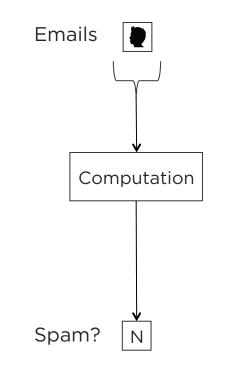


YAHOO!

ONLINE LINEAR CLASSIFICATION

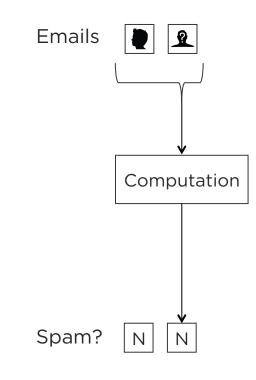






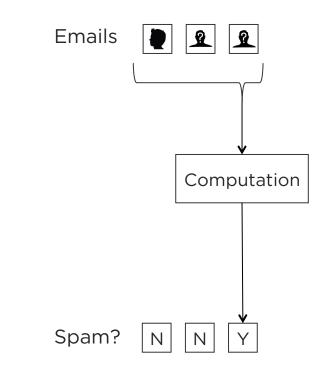


ONLINE MACHINE LEARNING



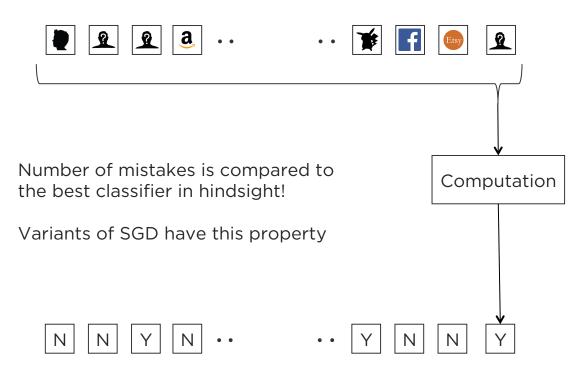


ONLINE MACHINE LEARNING





ONLINE MACHINE LEARNING



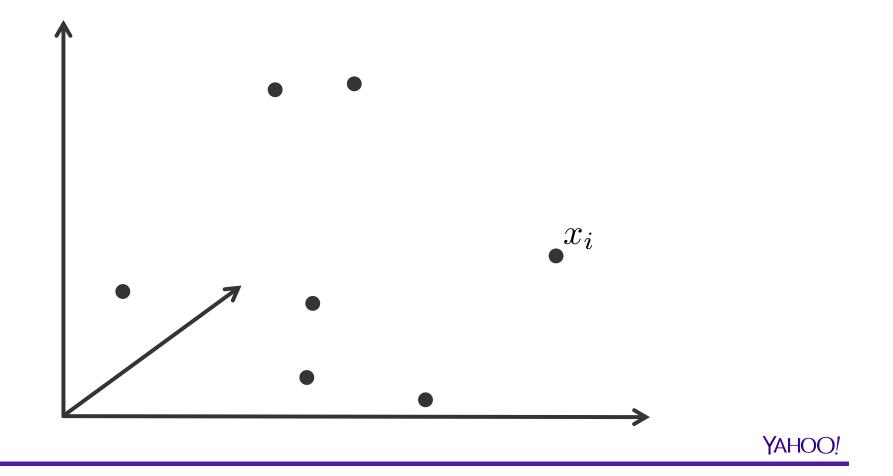
Prediction, Learning, and Games, Cesa-Bianchi, Lugosi, 2006

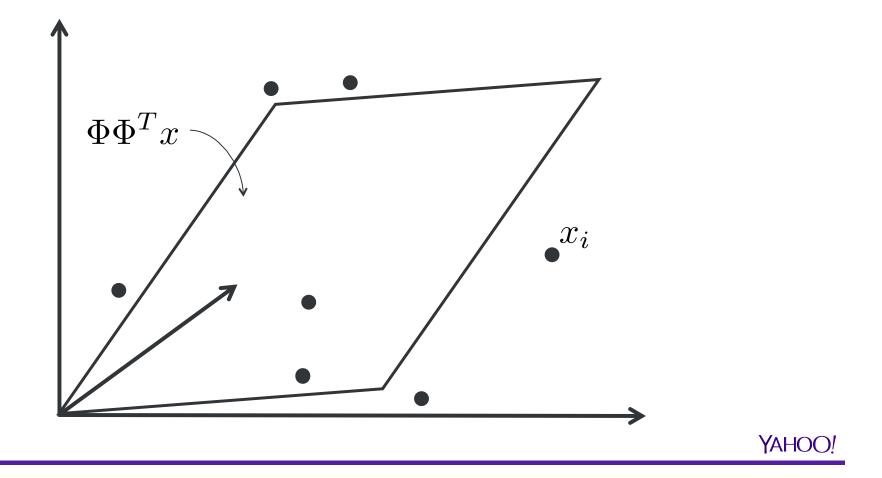


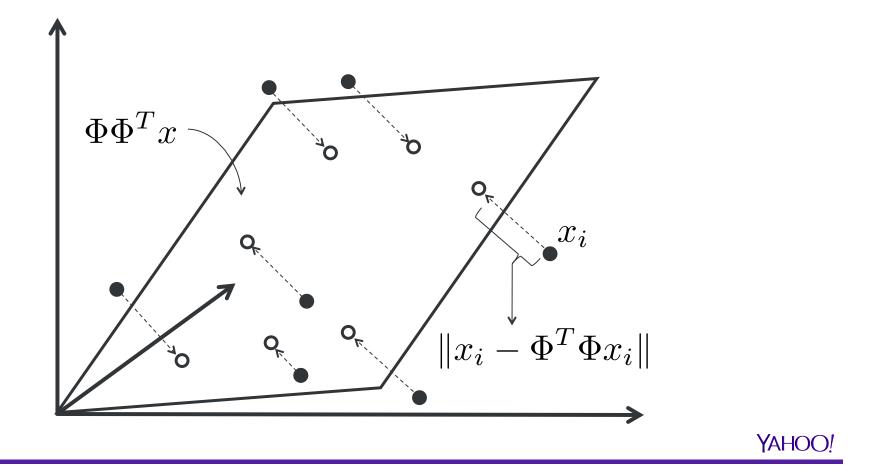
ONLINE PRINCIPAL COMPONENT ANALYSIS

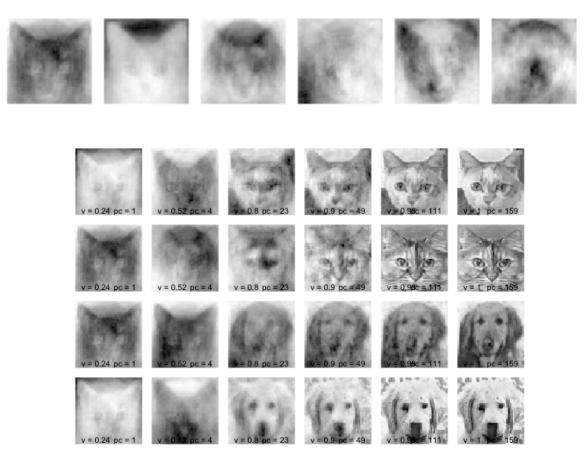
Online Principal Components Analysis, Boutsidis, Garber, Karnin, Liberty 2014 Online PCA with Spectral Bounds, Karnin, Liberty, 2015











Eigenpets: https://bioramble.wordpress.com/2015/09/01/



ONLINE PRINCIPAL COMPONENT ANALYSIS

Algorithm 1 Fixed Error: Conceptual Algorithminput: X, Δ $U \leftarrow$ all zeros matrixfor $x_t \in X$ doif $||(I - UU^T)X_{1:t}||^2 \ge \Delta$ Add the top left singular vector of $(I - UU^T)X_{1:t}$ to Uyield $y_t = U^T x_t$ end for

Online PCA with Spectral Bounds, Karnin, Liberty, 2015



Online PCA with Spectral Bounds Online PCA with Spectral Bounds



ONLINE K-MEANS CLUSTERING

An Algorithm for Online K-Means Clustering, Liberty, Sriharsha, Sviridenko, 2014

YAHOO!

K-MEANS CLUSTERING

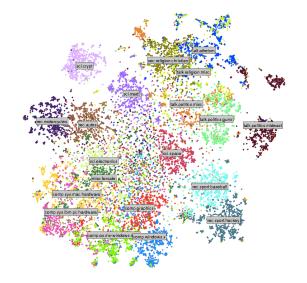
http://en.wikipedia.org/wiki/MNIST_database

http://research.ics.aalto.fi/mi/software/ne/



K-MEANS CLUSTERING

- Roughly 20,000 documents
- 20 topics:
 - Graphics
 - PC hardware
 - Baseball
 - For-sale
 - Politics
 - ...



http://gwone.com/~jason/20Newsgroups/

http://research.ics.aalto.fi/mi/software/ne/



K-MEANS CLUSTERING

1) One can cluster points fully online

2) Create only slightly more than k centers

3) Be competitive with the best offline clustering to k clusters

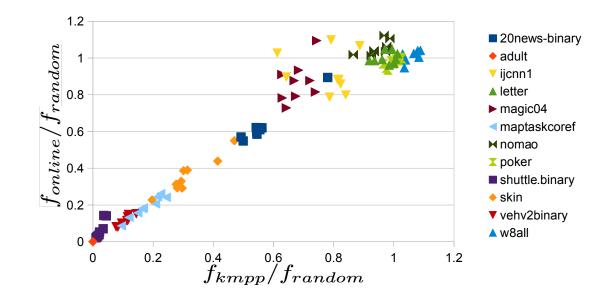
Algorithm 2 Online k-means algorithm

input: V, k $C \leftarrow \text{first } k + 1 \text{ distinct vectors in } V; \text{ and } n = k + 1$ (For each of these yield itself as its center) $w^* \leftarrow \min_{v,v' \in C} \|v - v'\|^2/2$ $r \leftarrow 1; q_1 \leftarrow 0; f_1 = w^*/k$ for $v \in$ the remainder of V do $n \leftarrow n + 1$ with probability $p = \min(D^2(v, C)/f_r, 1)$ $C \leftarrow C \cup \{v\}; q_r \leftarrow q_r + 1$ if $q_r \ge 3k(1 + \log(n))$ then $r \leftarrow r + 1; q_r \leftarrow 0; f_r \leftarrow 2 \cdot f_{r-1}$ end if yield: $c = \arg\min_{c \in C} \|v - c\|^2$ end for

An Algorithm for Online K-Means Clustering, Liberty, Sriharsha, Sviridenko 2015



ONLINE K-MEANS CLUSTERING



An Algorithm for Online K-Means Clustering, Liberty, Sriharsha, Sviridenko 2015 k-means++: the advantages of careful seeding, Arthur, Vassilvitskii, 2006



STREAMING ALGORITHMS

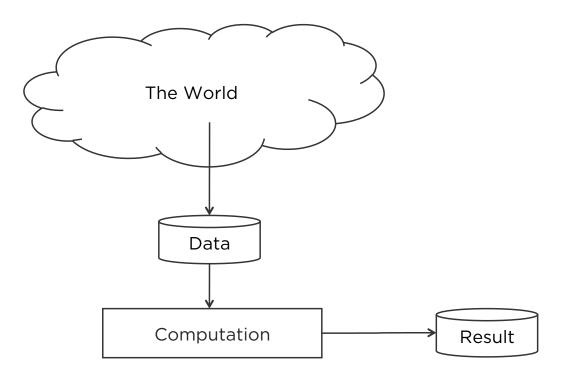
OPEN SOURCE FROM YAHOO EDO LIBERTY



DATASKETCHES.GITHUB.IO

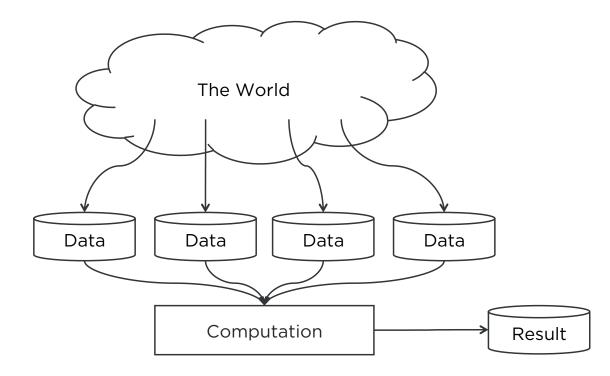






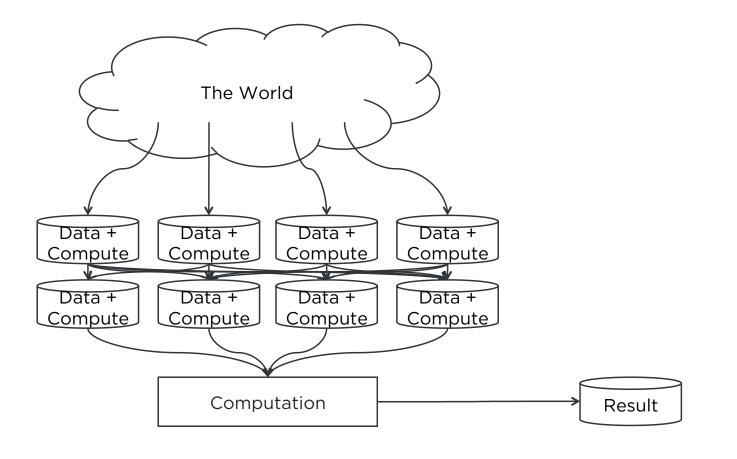


DISTRIBUTED STORAGE



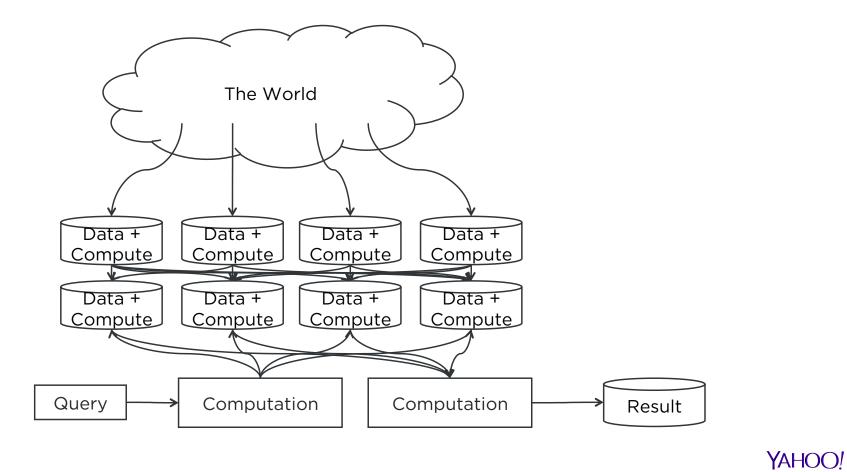


DISTRIBUTED MODEL (MAP/REDUCE, MESSAGE PASSING, ...)

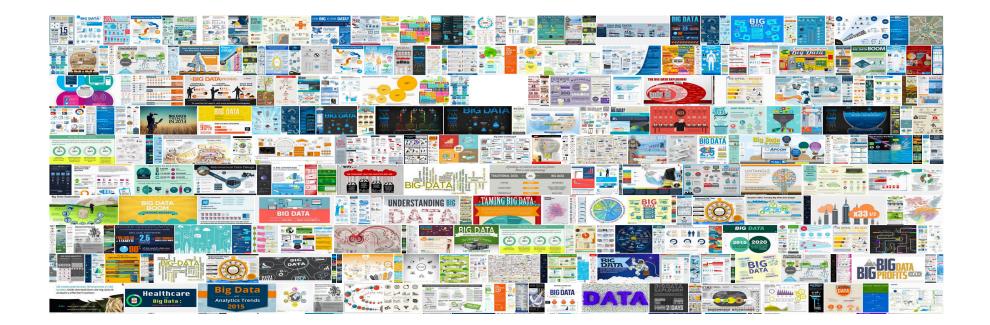




DISTRIBUTED MODEL (INDEXES, TABLES, DATABASES, ...)

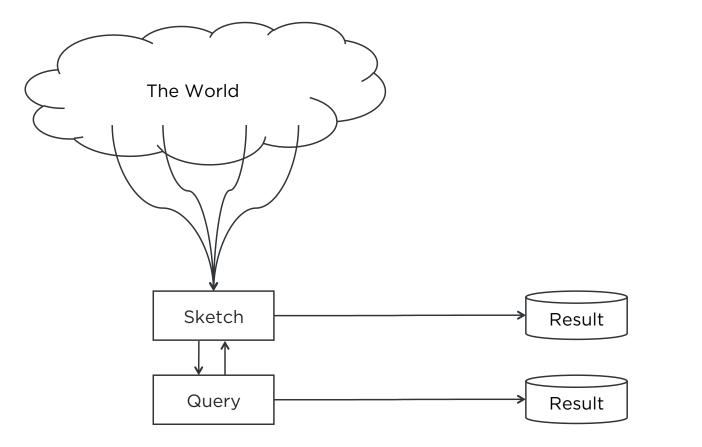


BIG-DATA *META* **INFOGRAPHIC**

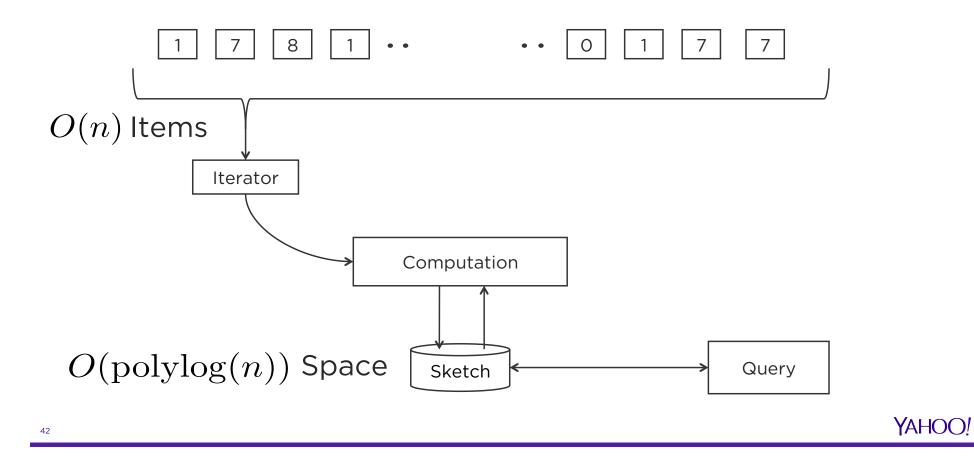




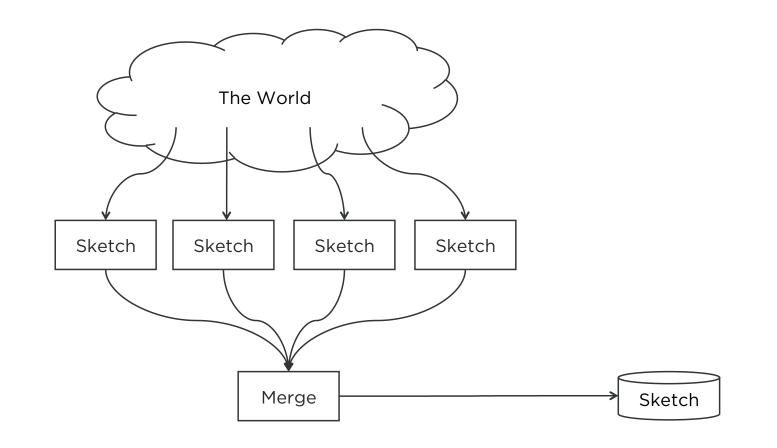
THE STREAMING COMPUTATIONAL MODEL



THE STREAMING COMPUTATIONAL MODEL

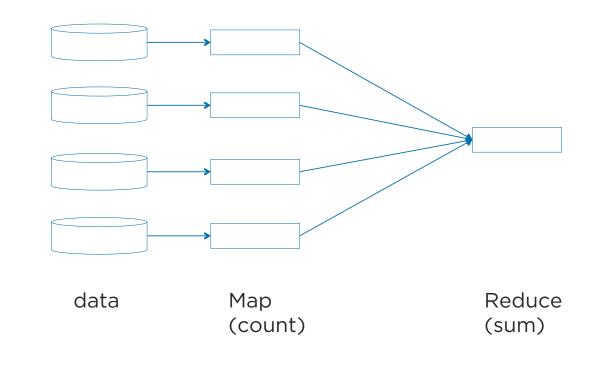


THE DISTRIBUTED STREAMING COMPUTATIONAL MODEL





Number of users (easy)





Web Site Logs

Time	User ID	Site	Time Spent Sec	ltems Viewed
9:00	U1	Apps	59	5
9:30	U2	Apps	179	15
10:00	U3	Music	29	3
1:00	U1	Music	89	10

Unique User Queries

- Unique users viewing Apps since 9:45...?
- Unique users visiting Apps site AND Music site?
- Unique users visiting Apps site AND NOT Music site?

Quantile Queries

- The median and 95%ile Time Spent seconds by ...?
- A Frequency Histogram of Time Spent by Split-Points specified at query time?

Financial Transactions System Log

Time	User ID	Site	Purchased	Revenue
9:00	U1	Apps	FaceTune	\$3.99
9:30	U2	Apps	Minecraft	\$6.99
10:00	U3	Music	Purple Rain	\$1.29
10:05	U3	Apps	Minecraft	\$6.99

Frequency Queries

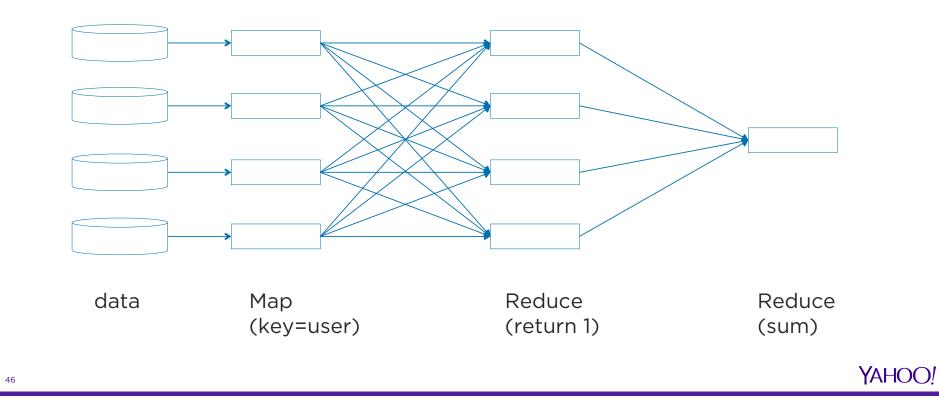
• The numbers of times each app was purchased

Join Queries

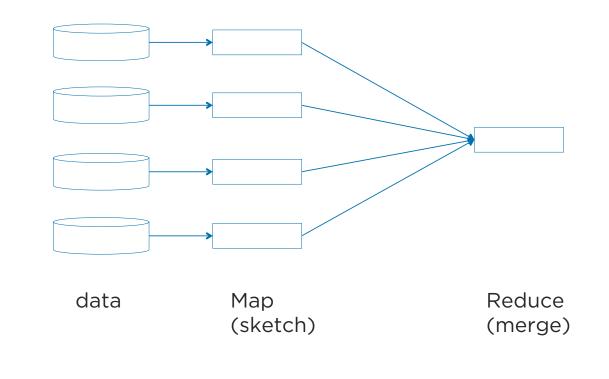
• For all users that purchased Apps, what is the average / median time spent?



Number of <u>unique</u> users (hard)



Number of <u>unique</u> users (made easy)





Current Sketch Implementations

Count Unique Sketches

- Both Theta Sketches* and HLL Sketches
- Estimating Cardinality of a stream of identifiers with duplicates
- Set Operations (e.g., Union, Intersection, and Difference)
- Can be extended to produce approximate Joins

Quantiles Sketches

 Normal or Inverse PMF's, CDF's of streams of numeric values, using after-the-fact queries.

Frequent Item Sketches

- Identify the Heavy Hitters of arbitrary objects from a stream of objects
- Estimate the frequency of any item from the stream



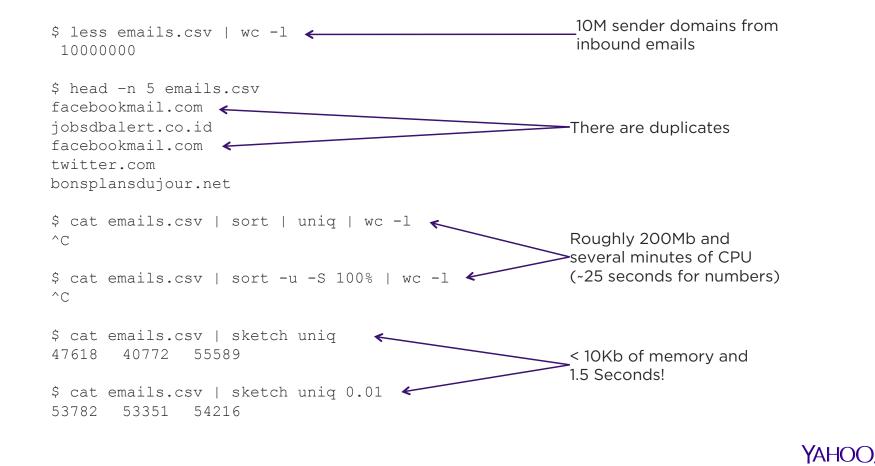
DataSketches.GitHub.io Open Source Library

- Dedicated to production quality Sketch implementations.
 - These are not toy algorithms!
 - Heavily used within Yahoo
- Common Attributes
 - True streaming. Single pass, "one-touch" algorithms for either *real-time* or *batch*
 - All Sketches are Mergeable, which makes them highly parallelizable.
 - Designed for multiple large-scale computing environments:
 - Core of library is coded in Java with no external dependencies
 - Easy integration into virtually any system environment
 - Adaptors for Hadoop/Pig and Hadoop/Hive environments
 - Standard library promotes sharing across platforms and organizations
 - Maven deployable and registered with Maven Central Repository
 - http://search.maven.org/#search/ga/1/datasketches
 - Comprehensive unit tests and testing tools are provided
 - Extensive documentation with Systems Developers in mind
 - All algorithms are backed by published mathematical theory



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Counting distinct elements example



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