

Using Big Data to Improve Product Quality

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Outline

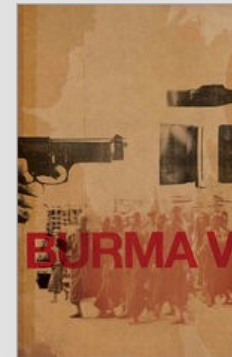
- Netflix recommendations
- What is big data?
- Industry examples
- A technical example: singular value decomposition for recommender systems
- Thought exercises

NETFLIX

“Connecting people to the movies they love”

Critically-acclaimed Fight-the-System Documentaries

Based on your interest in...

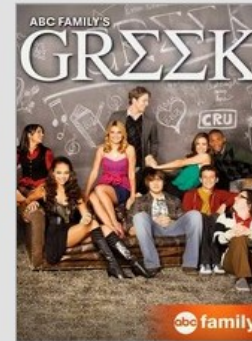
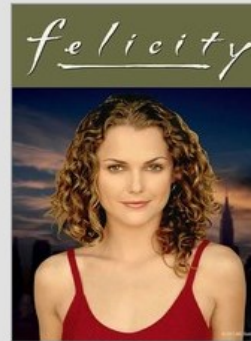


Teen TV Dramas Featuring a Strong Female Lead

Your taste preferences created this row.

TV Dramas
Strong Women.

As well as your interest in...



Movies Featuring an Epic Nicolas Cage Meltdown



<http://genresofnetflix.tumblr.com/>

Movie Received

[Browse](#)**Contempt**

Rate this title: ☆☆☆☆☆

Click one of the stars above to rate this movie. Rate movies you've seen to get personalized recommendations based on your ratings.



No Opinion

(click the stars)

Rate Your Recent Return

To Reveal 2 Movies You'll Love

1

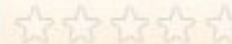
Add



Not Interested

2

Add



Not Interested

Intuitions

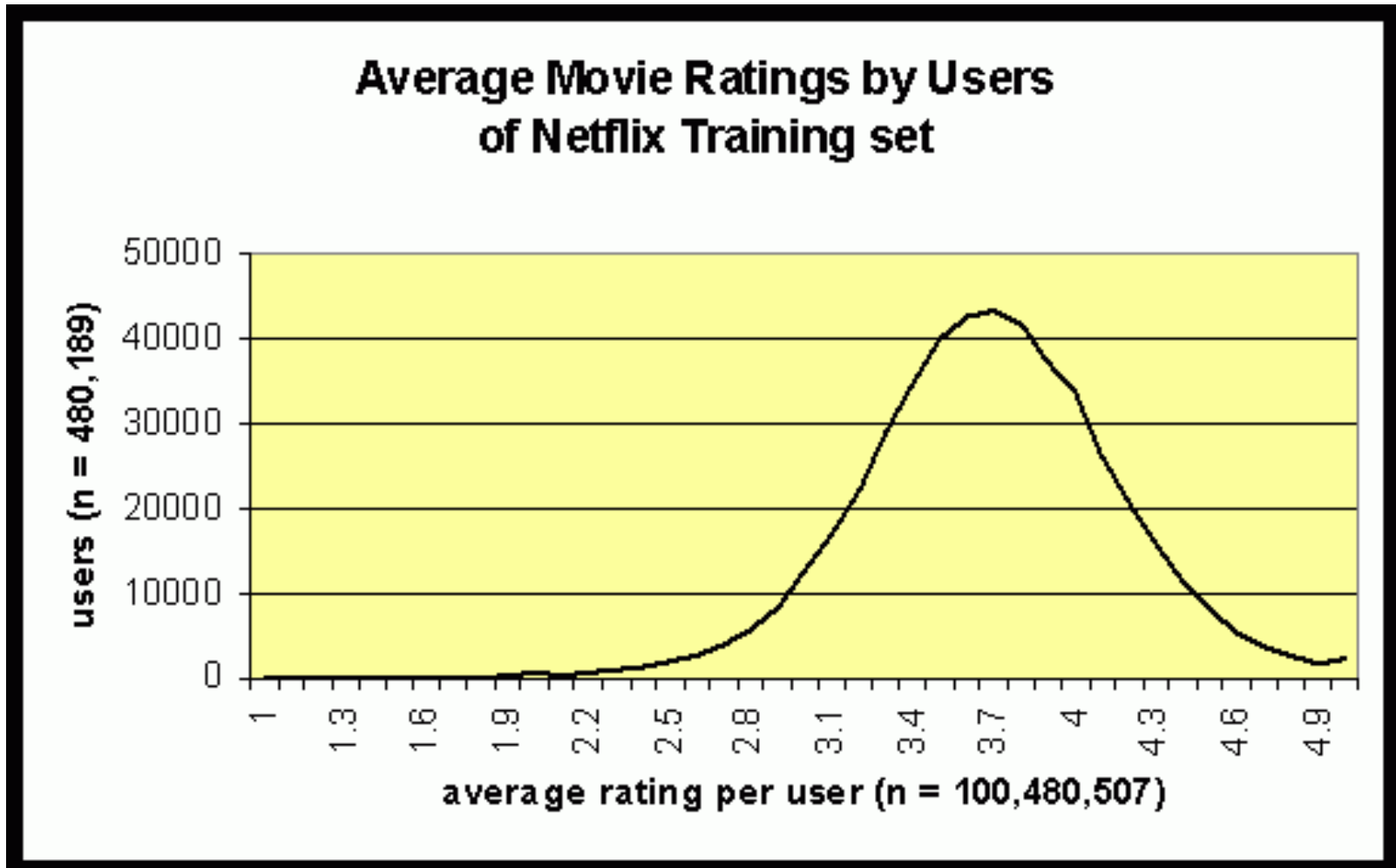
1- People rate movies high

*They **generally** don't bother rating bad movies*

I ONLY RATE
BAD MOVIES!



People rate movies high



<http://www.netflixprize.com/community/viewtopic.php?pid=5941#p5941>

Intuitions

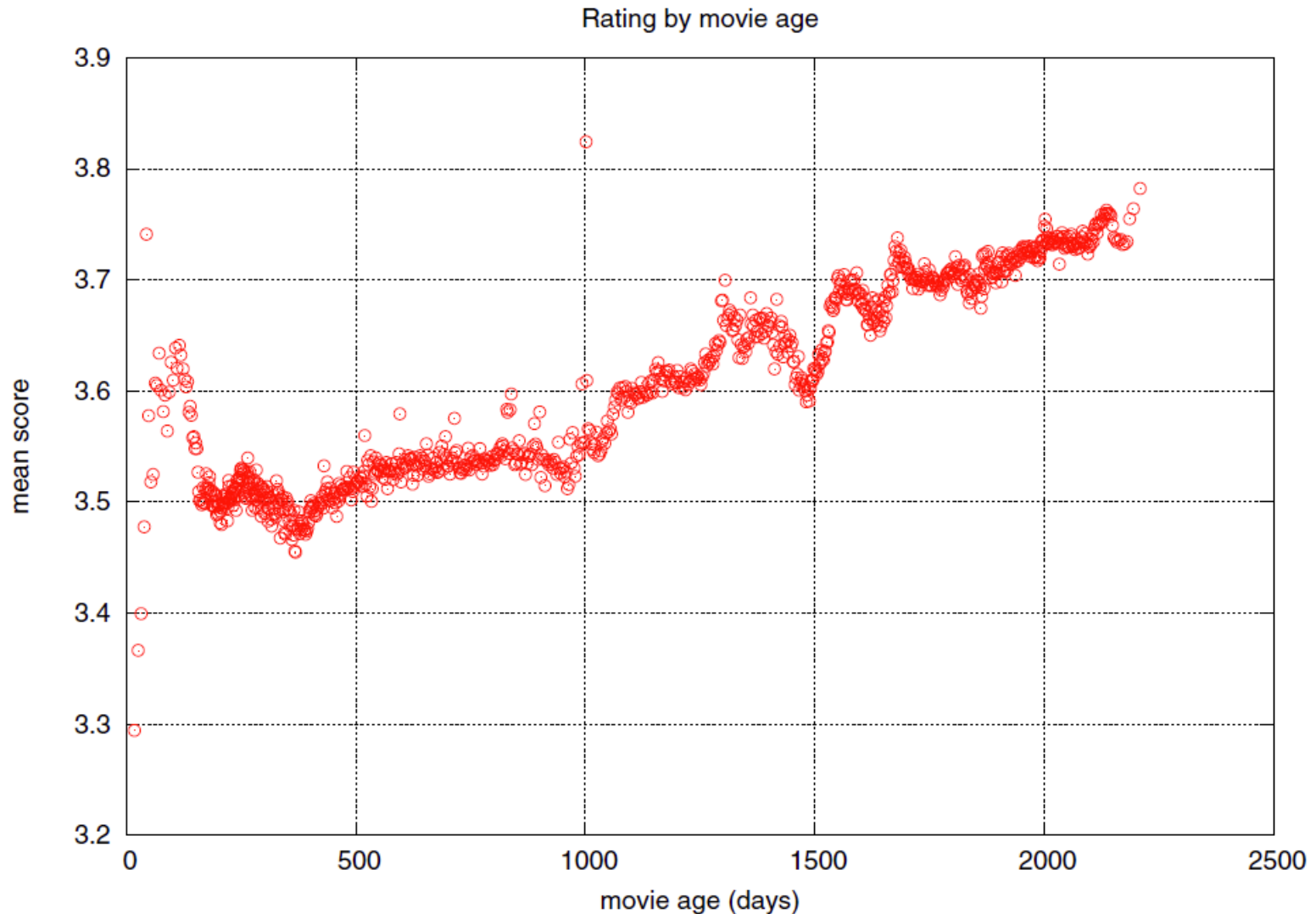
2- Older movies are rated higher

Nostalgia

Netflix keeps only the best of old movies



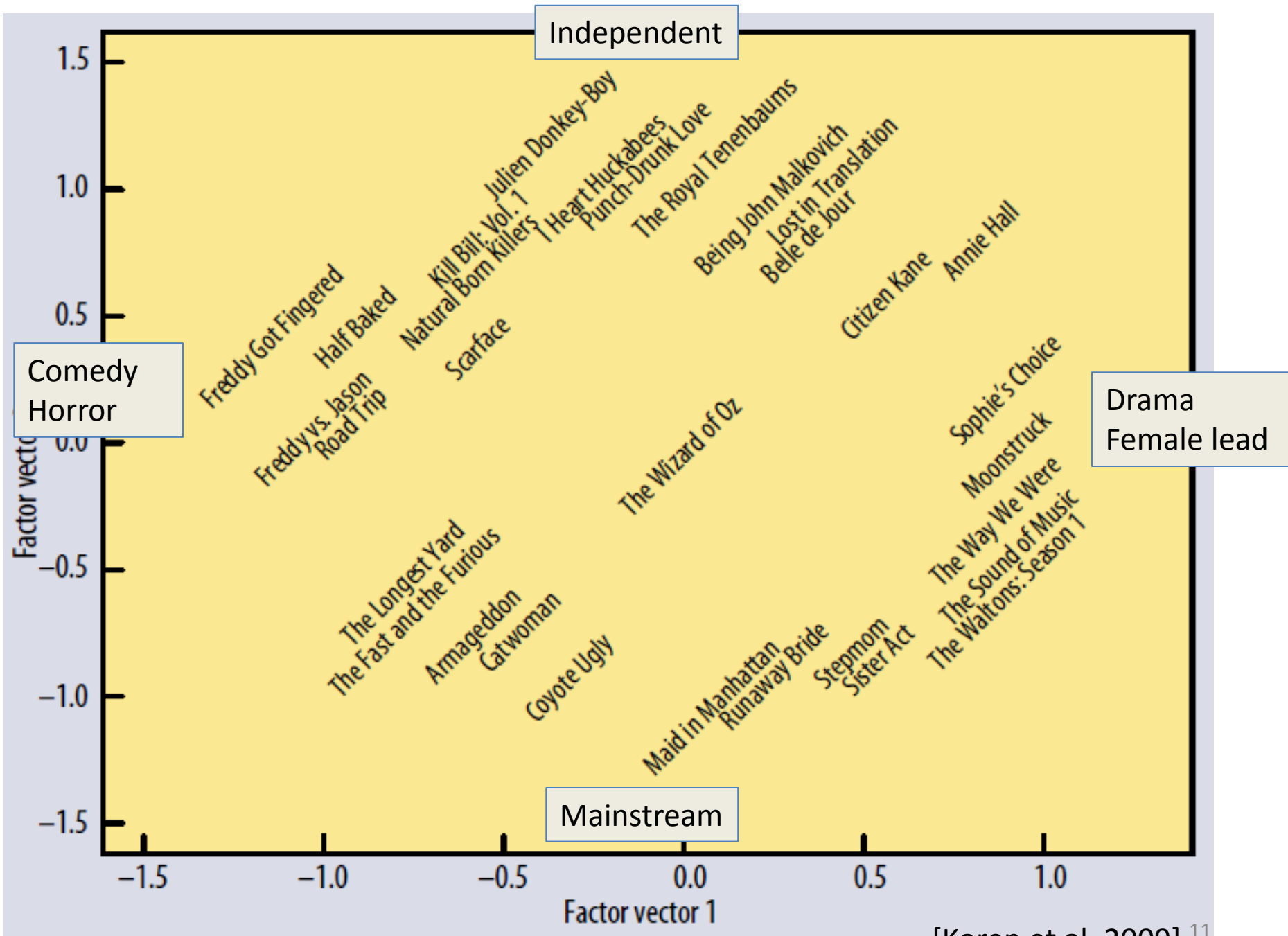
Older movies are rated higher



Intuitions



3- Movies have genres



Take-away

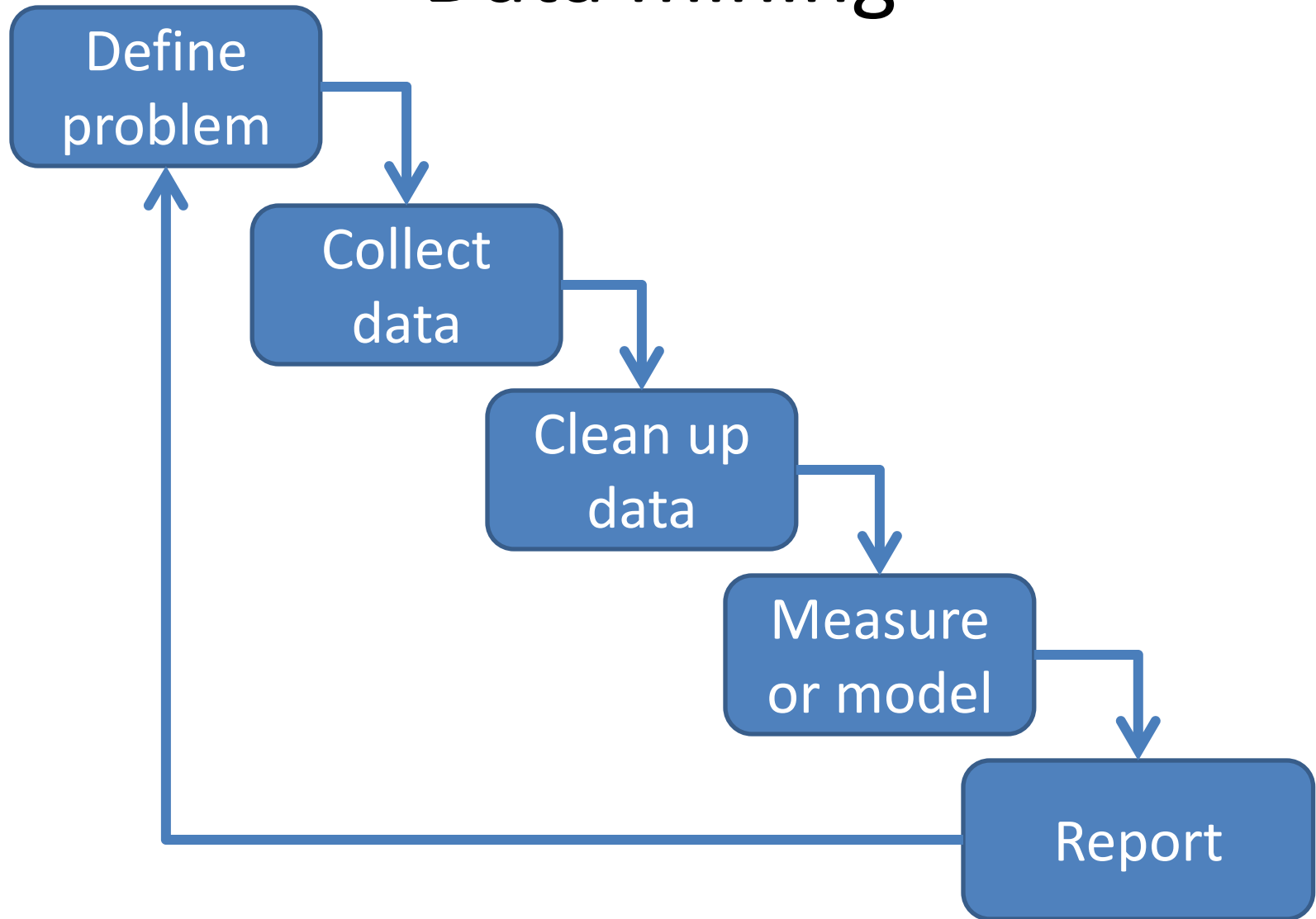
- Confirm or reject intuitions with data

WHAT IS BIG DATA?

BIG DATA

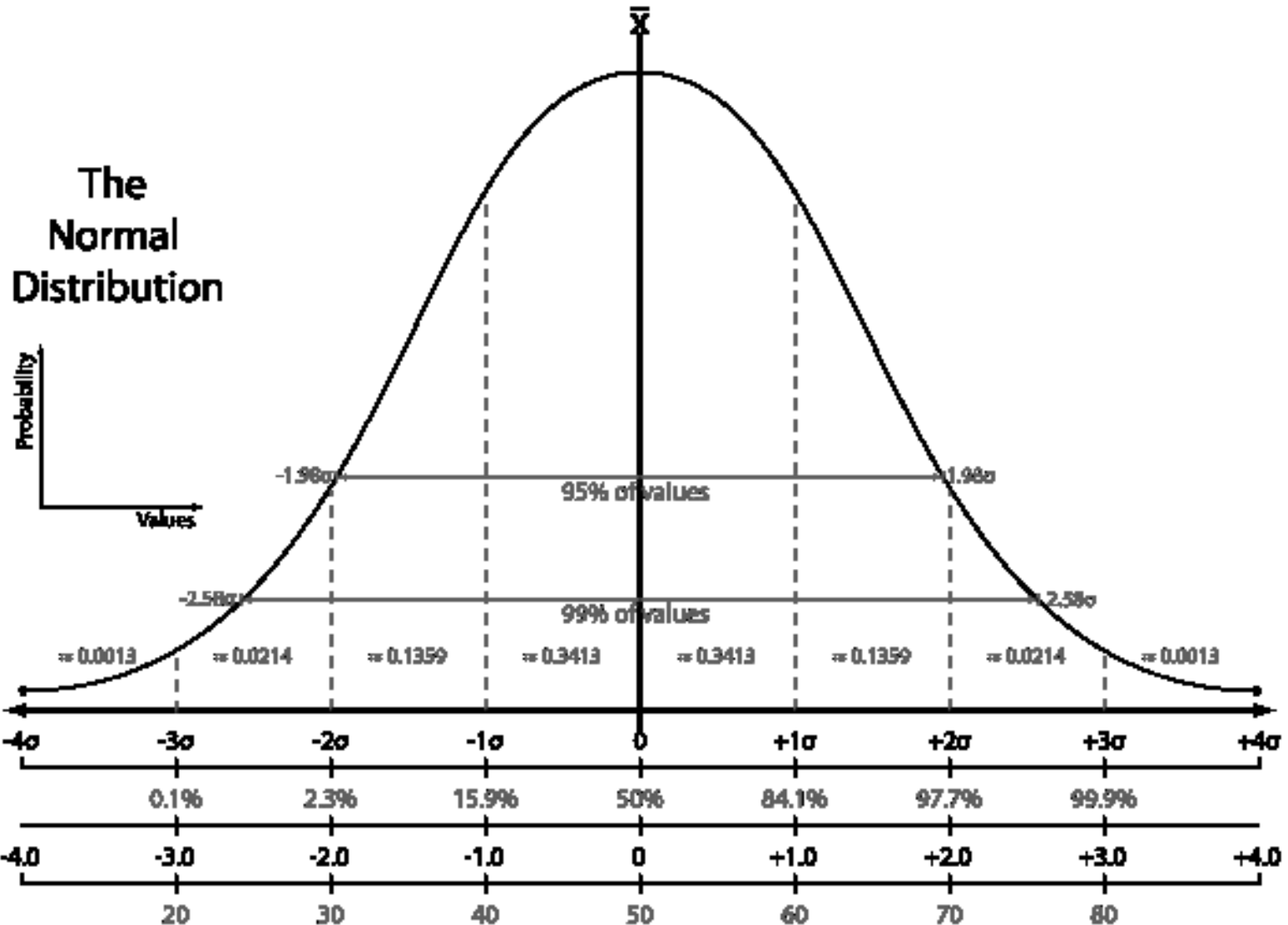


Data mining



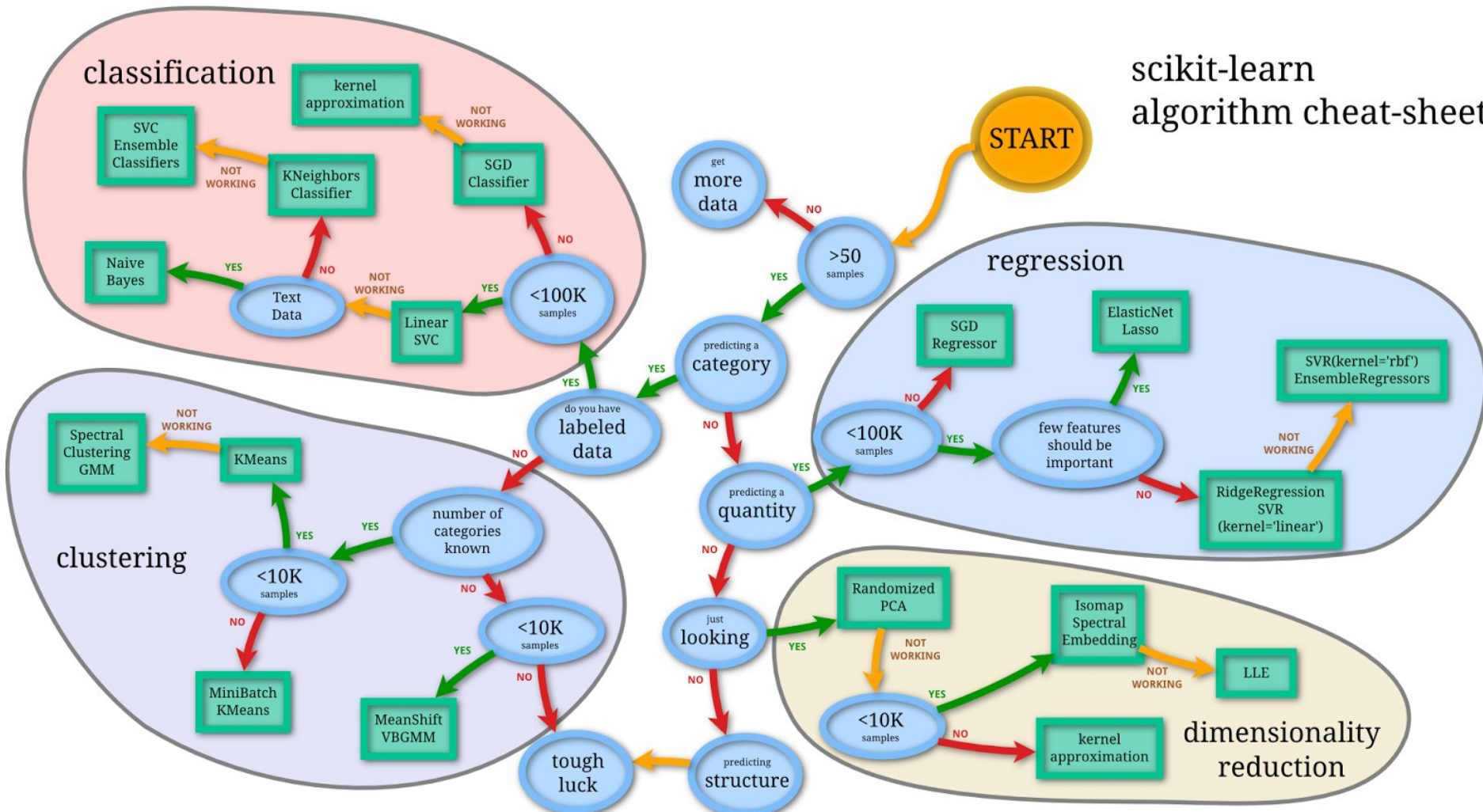
Statistics

The Normal Distribution



Machine learning

scikit-learn
algorithm cheat-sheet

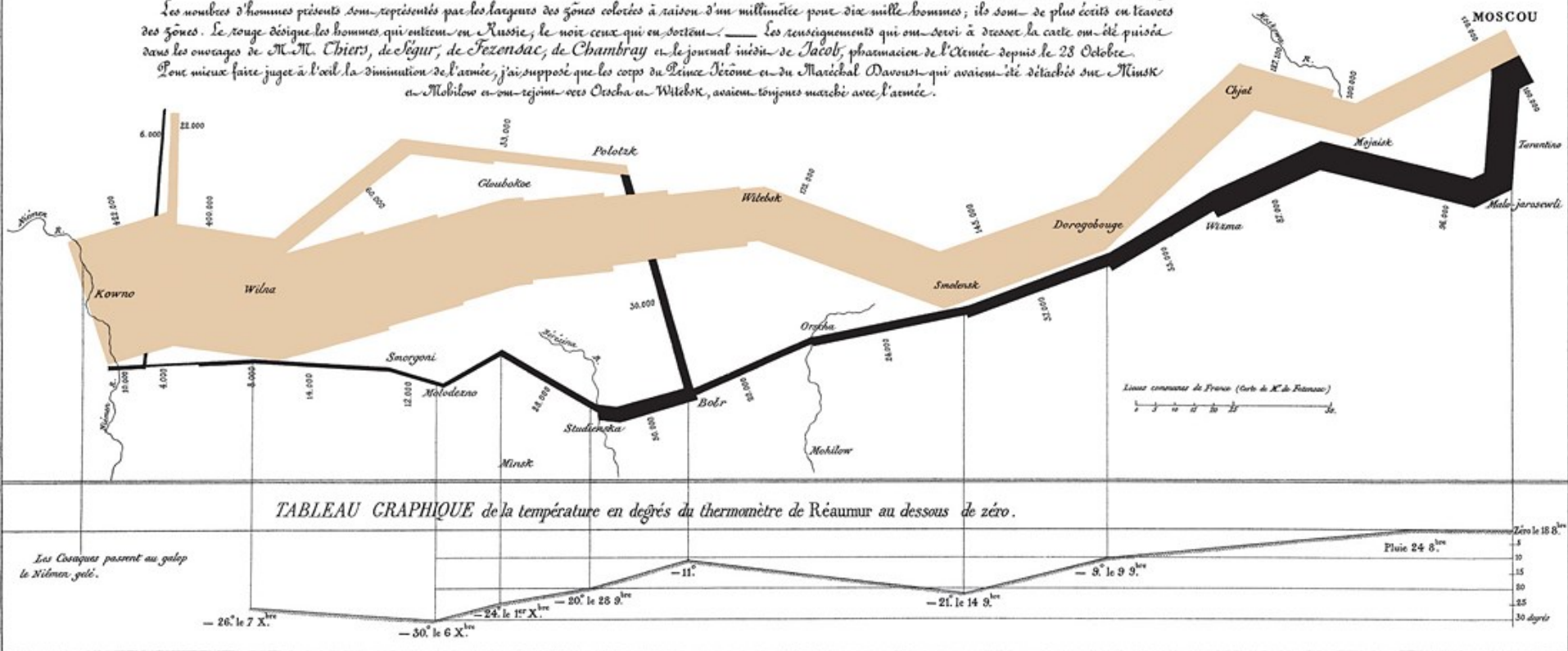


<http://n-chandra.blogspot.com/2013/01/picking-machine-learning-algorithm.html>

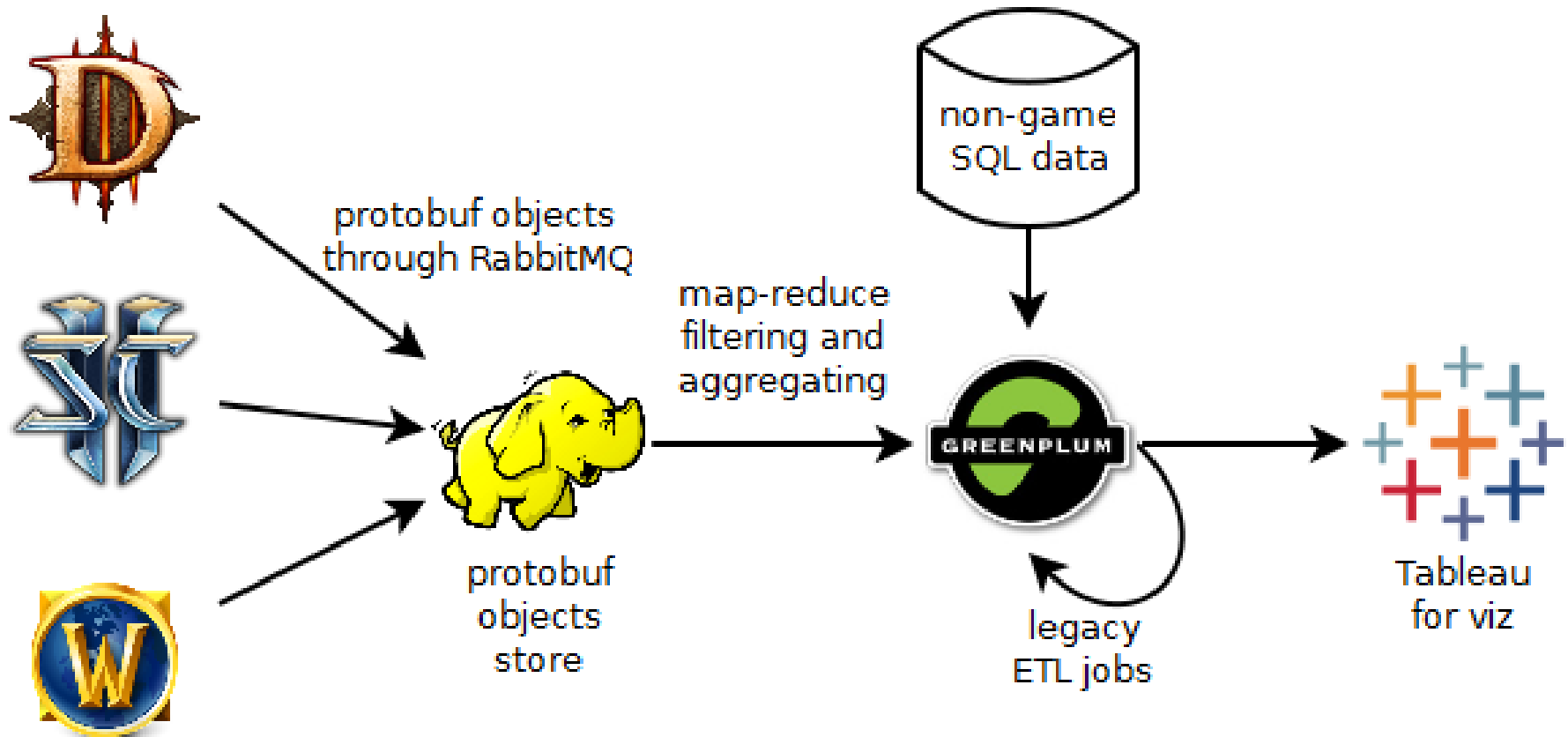
Visualization

Carte Figurative des pertes successives en hommes de l'Armée Française dans la campagne de Russie 1812-1813.
Dressée par M. Minard, Inspecteur Général des Ponts et Chaussées en retraite Paris, le 20 Novembre 1869.

Les nombres d'hommes présents sont représentés par les largeurs des zones colorées à raison d'un millimètre pour dix mille hommes; ils sont de plus écrits en lettres des zones. Le rouge désigne les hommes qui ont été en Russie, le noir ceux qui en sont sortis. Les renseignements qui ont servi à dresser la carte ont été puisés dans les ouvrages de M. M. Thiers, de Ségur, de Fozensac, de Chambray et le journal inédit de Jacob, pharmacien de l'Armée depuis le 23 Octobre. Pour mieux faire juger à l'œil la diminution de l'armée, j'ai supposé que les corps du Prince Jérôme et du Maréchal Davoust qui avaient été détachés sur Minsk et Mohilow et qui rejoignent vers Otscha et Wiltschok, avaient toujours marché avec l'armée.



Data engineering



<http://bowling-bash.blogspot.com/2013/10/blizzards-hadoop-platform.html>

Take-aways

- Many tools and processes
- Some old, some recent
- Pick the right one for the problem at hand

INDUSTRY EXAMPLES

Google



FredMeyer®



T U R N S T **Y** L E



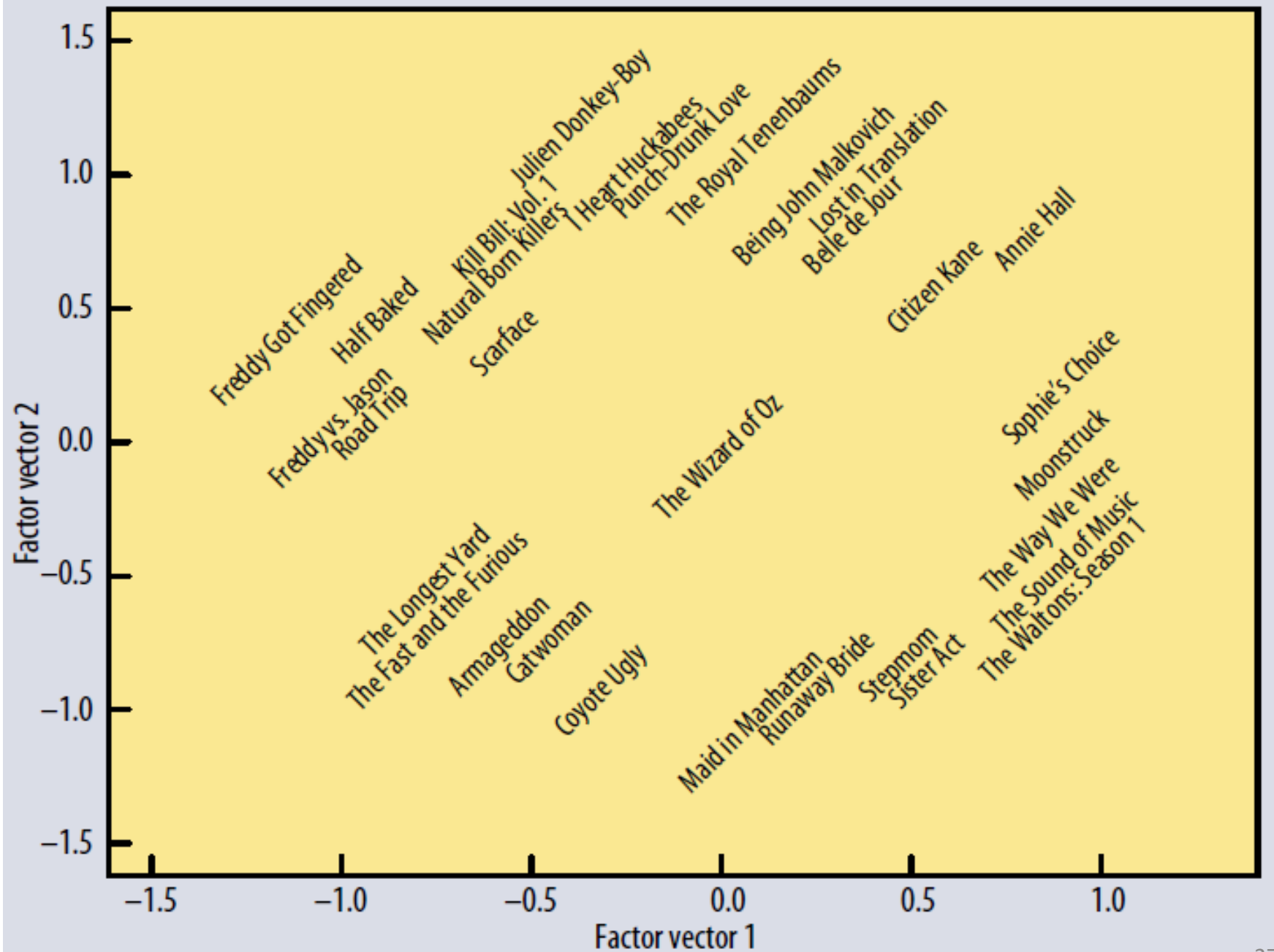
amazon.com[®]

NETFLIX

Take-aways

- Useful
- Everywhere
- Creepy?

SINGULAR VALUE DECOMPOSITION



Ratings data

	The Longest Yard	Citizen Kane	Fast & Furious	Kill Bill	Sound of Music
Alice	1	4	1		5
Bob	5		4	4	
Claire		4	2		4
Dan	4	1		3	
Eve		3			5
Frank	3	3	5	2	

(usually 99% sparse)

Singular Value Decomposition

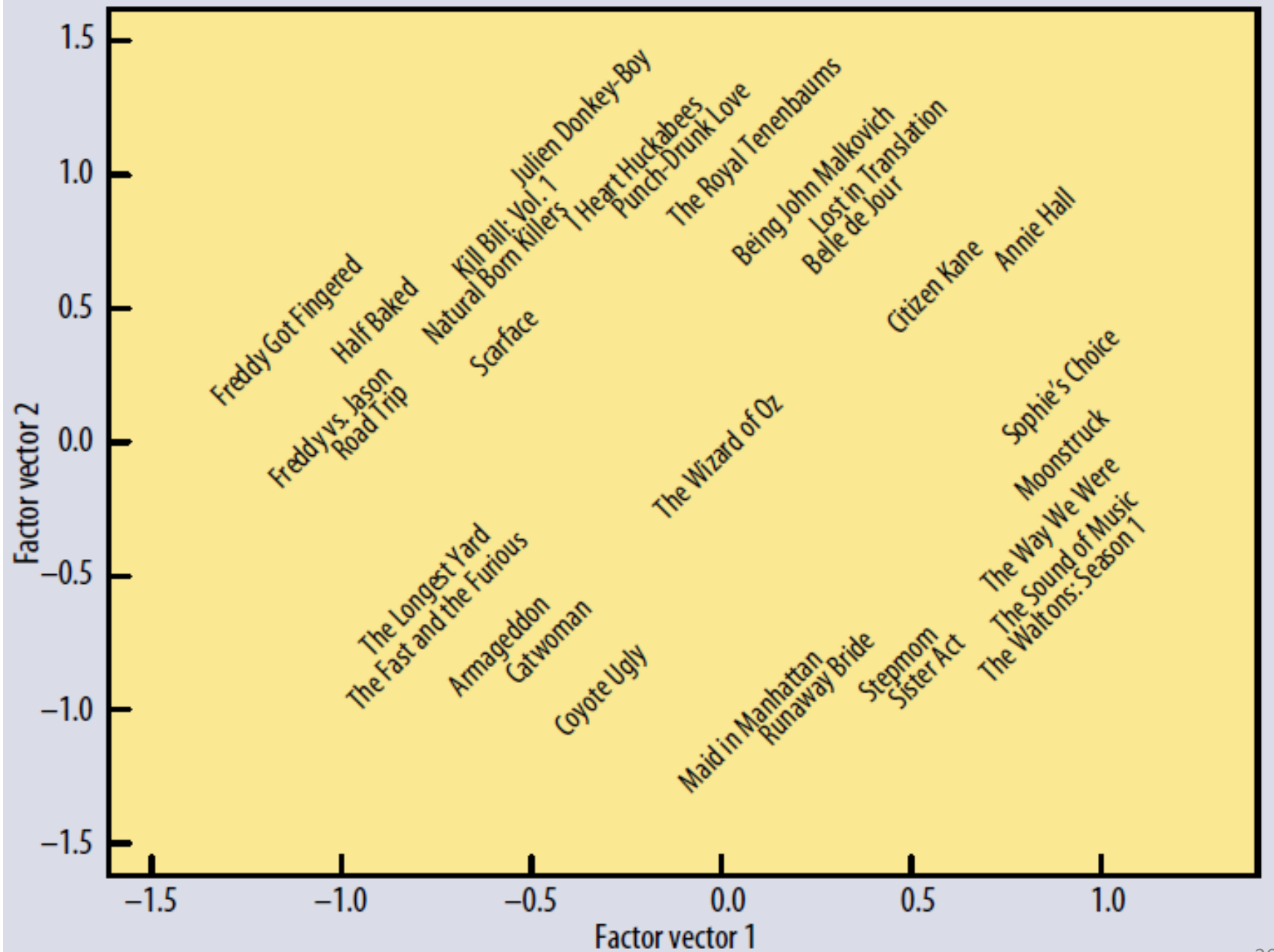
						Users				Tastes							Movies					
	The Longest Yard	Citizen Kane	Fast & Furious	Kill Bill	Sound of Music	=		*0.1	Factor 1	Factor 2	X		F1	F2	X		*0.1	The Longest Yard	Citizen Kane	Fast & Furious	Kill Bill	Sound of Music
Alice	1	4	1	M	5		Alice	-4	7	F1		18	0	F1		-4	-4	-4	-4	-5		
Bob	5	M	4	4	M		Bob	-5	-4	F2		0	4	F2		-5	4	-6	1	4		
Claire	M	4	2	M	4		Claire	-4	2													
Dan	4	1	M	3	M		Dan	-4	-5													
Eve	M	3	M	M	5		Eve	-4	0													
Frank	3	3	5	2	M		Frank	-4	0													

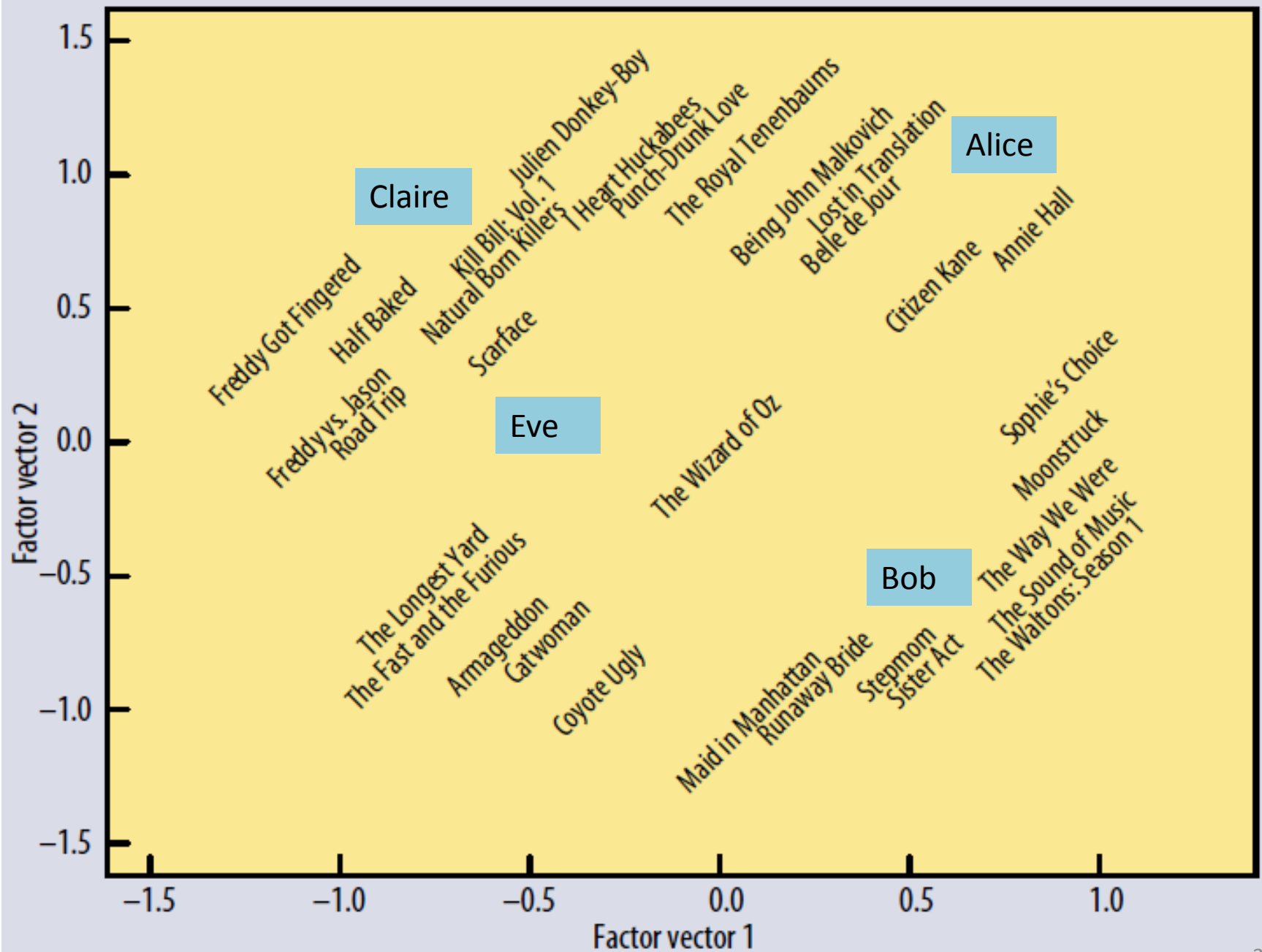
=

X

X

M = global average = 3.3





Densifying sparse matrices

Number of ratings x100

- Storage x100
- SVD is $O(N^3)$, CPU x1M

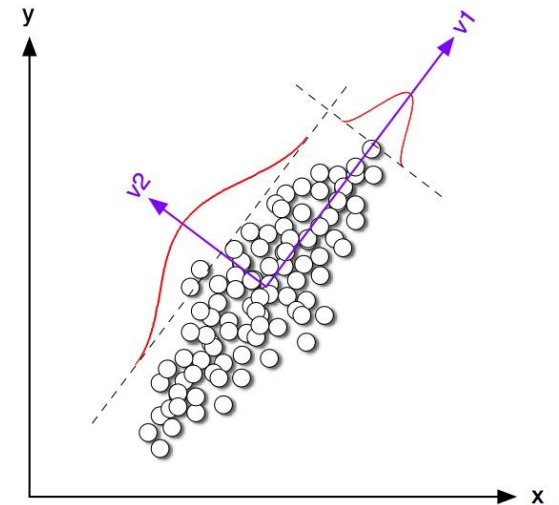
Minimize square error on known ratings using stochastic gradient descent

$$\min_{q^*, p^*} \sum_{(u,i) \in K} (r_{ui} - q_i^T p_u)^2 + \lambda (\|q_i\|^2 + \|p_u\|^2)$$

	Armageddon	Citizen Kane	Fast & Furious	Kill Bill	Sound of Music
Alice	1	4	1	M	5
Bob	5	M	4	4	M
Claire	M	4	2	M	4
Dan	4	1	M	3	M
Eve	M	3	M	M	5
Frank	3	3	5	2	M

Similar techniques

- Latent Semantic Analysis
 - Movies -> documents, ratings -> tf-idf
- Latent Dirichlet Allocation
 - Cf “A theory of aspects as latent topics”
- Principal Component Analysis
 - Eigen value decomposition
 - Of the covariance matrix



Take-aways

- Matrices
- Need practical methods
- The math escalates quickly
 - But you may need it ...

HOW WOULD YOU DO ...

Amazon

- “Frequently bought together”
 - Association rule mining
- Customer clustering
 - SVD
- (Stock management
 - Poisson process?)

Title

Questions that may already have your answer

- 2 [Is is possible to code an app that can detect and connect to peers without a central servers? How?](#)
- 0 [is it possible peer to peer connection over wan without any server? 1](#)
- 0 [Python Error "Connection reset by peer" 10054](#)
- 0 [Web Audio API and WebRTC 1](#)
- 0 [can peer A connect to peer B if server is on peer B? 1](#)

B *I* |     |     |   

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"This module provides the basic infrastructure for writing asynchronous socket service clients and servers."

<http://docs.python.org/2/library/asyncore.html#module-asyncore>

If I want a non- client-server architecture, such as peer to peer, can ~~asynchat/asyncore~~ still do the job?

draft saved

"This module provides the basic infrastructure for writing asynchronous socket service clients and servers." <http://docs.python.org/2/library/asyncore.html#module-asyncore>

If I want a non- client-server architecture, such as peer to peer, can asynchat/asyncore still do the job?

Tags

[python](#) [asyncore](#)

Similar Questions

- [Detecting socket close with Python's asyncore and smtpd](#)
- [asyncore python hangs](#)
- [Sending data using asyncore doesn't work](#)
- [python asyncore not keeping up with high data rates](#)
- [asyncore server: Request resulting in "socket.error">: \[Errno 32\] Broken pipe"](#)
- [Asyncore not working properly with Tkinter GUI](#)
- [Python, Asyncore and forks](#)
- [Which Python async library would be best suited for my code? Asyncore? Twisted?](#)
- [python asyncore keep track of clients](#)
- [How do I connect to IRC through a SOCKS proxy using asyncore/asynchat?](#)
- [How to re-establish asyncore connection with server \(solved\)](#)
- [Redirect a method call to something with a file descriptor - asyncore](#)
- [Python asyncore - multiple SMPP PDU's in one TCP packet](#)

Dating website

- Match %
 - SVD on weighted questions
 - “Would you rather be weird or normal?” = 99% weird
 - “Sex before marriage?”
 - SVD on tf-idf from profile essays
 - Cosine distance in SVD space
- Actual information in a profile
 - Everybody loves travelling, but they say it differently
 - “I love travelling!” = “travel” = low tf-idf
 - “Have a passport!” = “passport” = high tf-idf
 - Tf-idf is not always appropriate!

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THANKS

References

- Padhraic Smyth's slides on Netflix recsys
http://www.ics.uci.edu/~smyth/courses/cs277/public_slides/recommender_systems_part2.pdf
- Koren et al. 2009: Matrix factorization techniques for recommender systems
- <http://www.theatlantic.com/technology/archive/2014/01/how-netflix-reverse-engineered-hollywood/282679/>
- <http://online.wsj.com/news/articles/SB10001424052702303453004579290632128929194>