

Statistical and Data Literacy in the Era of Big Data

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http://www.boozallen.com/datascience

Ever since we first explored our world...



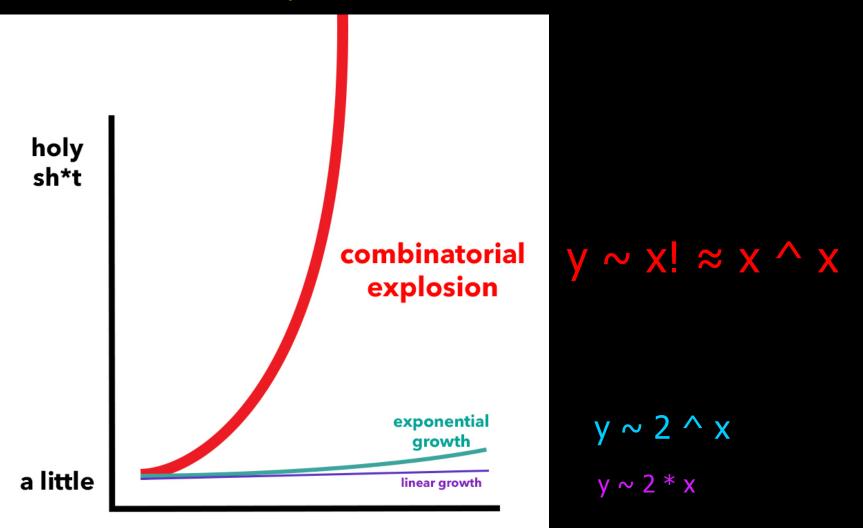
http://www.livescience.com/27663-seven-seas.html

...We have asked questions about everything around us.



https://jefflynchdev.wordpress.com/tag/adobe-photoshop-lightroom-3/page/5/

So, we have collected evidence (data) to answer our questions, which leads to more questions, which leads to more data collection, which leads to more questions, which leads to **BIG DATA!**



https://www.linkedin.com/pulse/exponential-growth-isnt-cool-combinatorial-tor-bair

Huge quantities of data are now being <u>used</u> everywhere!



https://datafloq.com/read/dont-let-big-data-hr-program-backfire/1445

Huge quantities of data are now being used everywhere!

"With great power comes great responsibility."– Spiderman's uncle (...or... Voltaire)



http://verix.com/6-new-articles-on-harnessing-the-power-of-big-data/

Data Ethics in Data Science in 2 parts: unbiased data & unbiased models

http://www.kirkborne.net/cds151/

EXCELLENT. WE CAN

MINING TECHNOLOGY

USE NON-LINEAR

TO OPTIMIZE OUR

RETAIL CHANNELS!

MATH AND DATA

Syndicate,

Feature

United

2000

1300

IF THAT'S THE

SAME THING AS

HAVING A GOOD

MEETING HERE.

http://dilbert.com/strip/2000-11-13

WE HAVE A GIGANTIC DATABASE FULL OF CUSTOMER BEHAVIOR INFORMATION.

The value of evaluation

scottadams@aol.con

www.dilbert.com

Data analysis can be fun and exploratory, BUT:

"If you torture the data long enough, it will confess to anything."

-Ronald Coase, economist

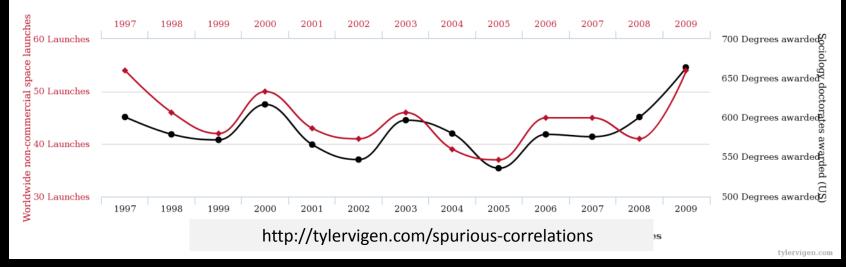
Statistical Fallacies can still appear in Data Science in the era of Big Data

Correlation ≠ Causation (beware hidden variables)

Worldwide non-commercial space launches

correlates with

Sociology doctorates awarded (US)



- Biased sampling or biased models (underfitting)
- Ignoring natural variance in the data (overfitting)
- Absence of Evidence ≠ Evidence of Absence

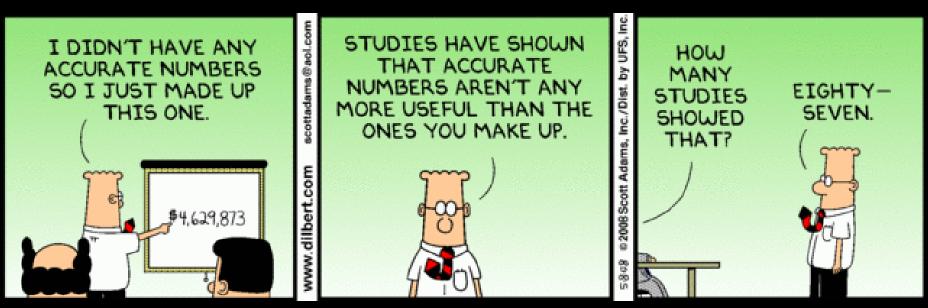
Quote from H.G. Wells (1903; writer) ...

"Statistical thinking will one day be as necessary for efficient citizenship as the ability to read and write."

Well, that day is here now! Statistical & Data Literacy Matters!

Quote from Steven Wright (comedian) ...

"42.7% of all statistics are made up on the spot."



http://dilbert.com/strip/2008-05-08

Quote from somebody (?) ...

"It is now beyond any doubt that cigarettes are the biggest cause of statistics"

a) "All models are wrong but some are useful"
b) "There are 3 types of lies – lies, damned lies, and statistics!"

c) "I am shocked that half the students in this country score below average on their standardized test scores"

d) "The best outcome for our education system is for more than half of our students to score below average on their standardized test scores!"

a) "All models are wrong but some are useful"
b) "There are 3 types of lies – lies, damned lies, and statistics!"

c) "I am shocked that half the students in this country score below average on their standardized test scores"

d) "The best outcome for our education system is for more than half of our students to score below average on their standardized test scores!"

a) "All models are wrong b George Box, famous statistician
b) "There are 3 types of lies – lies, damned lies, and statistics!" Benjamin Disraeli, British Prime Minister

c) "I am shocked that half the students in this country score below average on their standardized test <u>scores</u>" (famous American politician from 1990's)

d) "The best outcome for our education system is for more than half of our students to score below average on their standardized test scores!"

Let us examine these last two statements...

c) "I am shocked that half the students in this country score below average on their standardized test scores" (famous American politician from 1990's)

d) "The best outcome for our education system is for more than half of our students to score below average on their standardized test scores!"

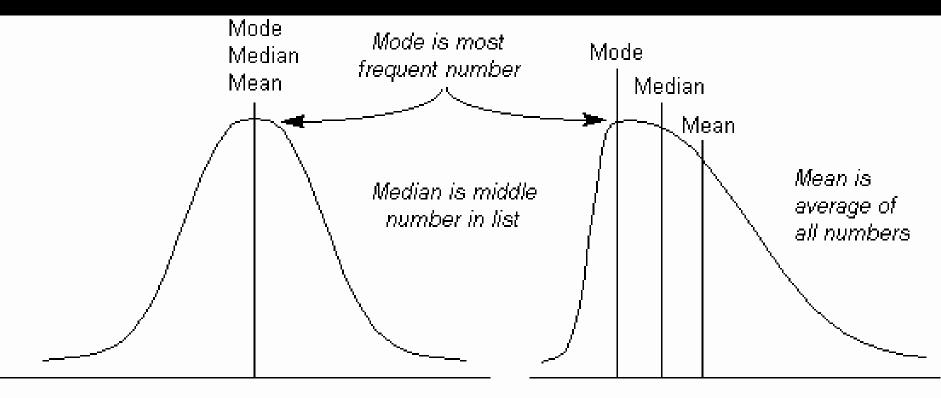
First question to ask ourselves:

What do we mean by *average*?

 $\bullet \bullet \bullet$

Is it the Mean or Median or Mode?

Means, Medians, and Modes, oh my



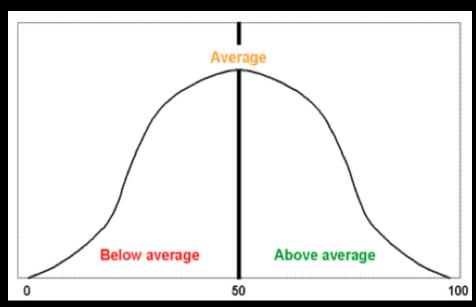
asymmetrical distribution

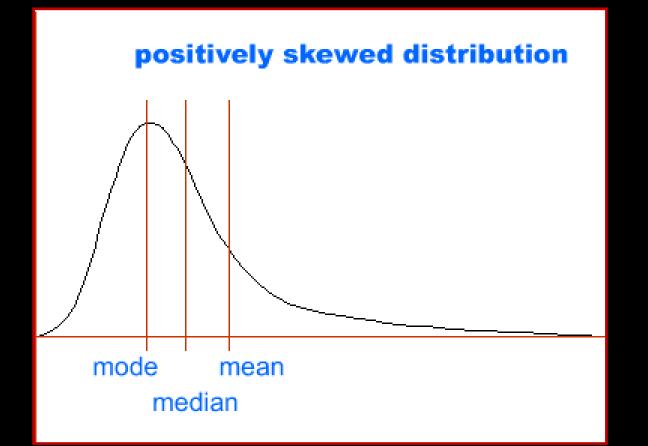
symmetrical distribution

Here is the analysis ...

"...half of the students in this country score below average on their standardized tests."

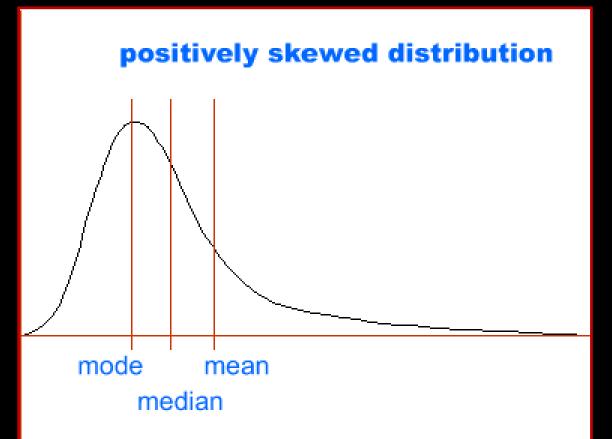
For a bell curve (normal) distribution of test scores, this statement is <u>neither</u> a good thing nor a bad thing, since half of the students will always score below average (by any definition of the word "average").





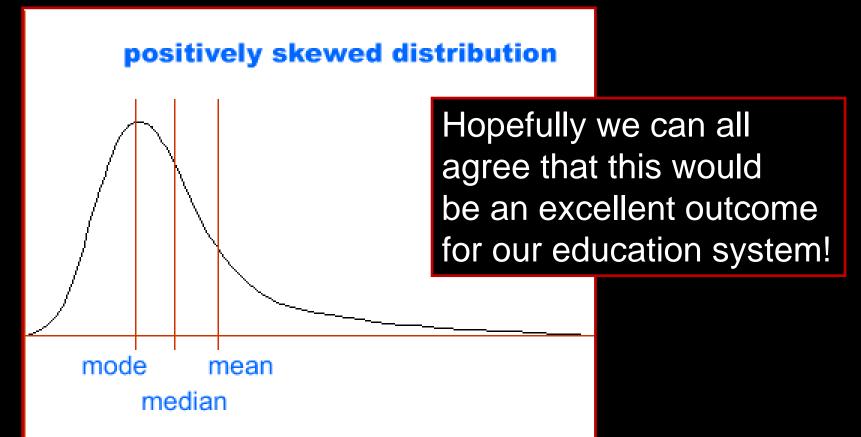
• This case corresponds to the statement:

"More than half of the values are below average" (i.e., below the mean)



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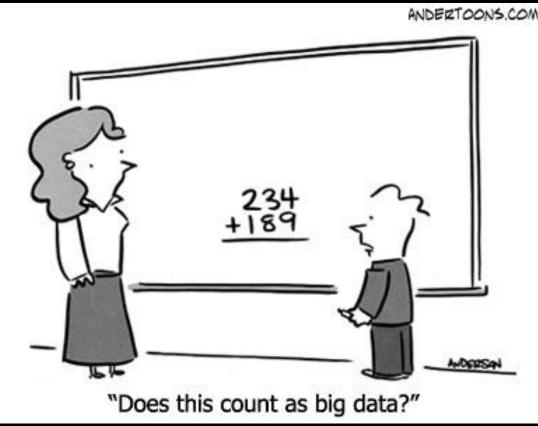
• This case corresponds to the statement:

"More than half of the values are below

Data Literacy Matters! be an excellent outcome for our education system!

Enough with the quiz!

- These were examples of statistical literacy
- Data Literacy includes statistical literacy ... and more...

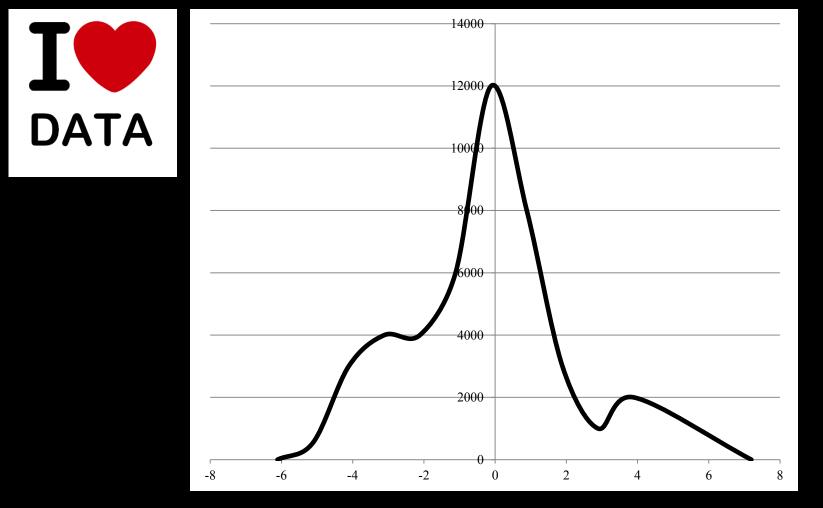


The 🕱 6 Commandments of Data Science

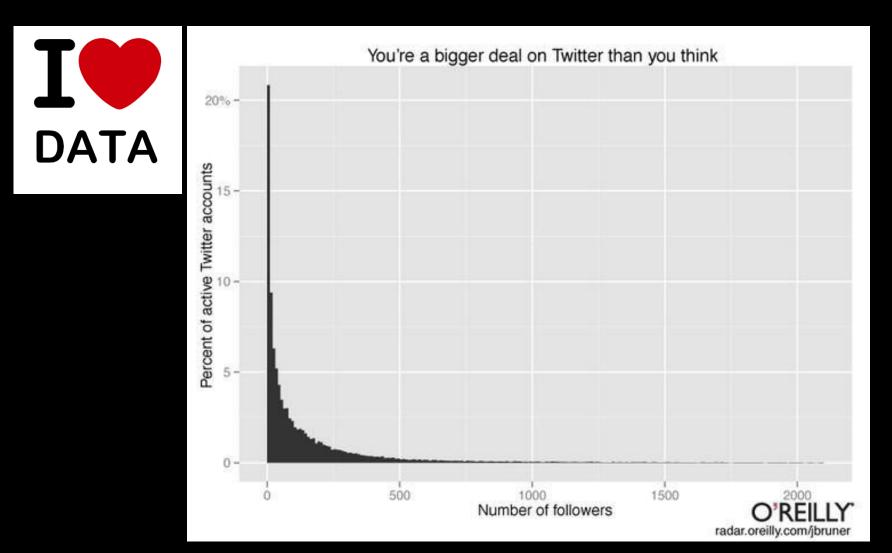
http://www.statisticsviews.com/details/feature/5459931/Five-Fundamental-Concepts-of-Data-Science.html

- **1.** Begin with the end in mind
- 2. Data Science is Science
- 3. Know thy data
- 4. Love thy data
- 5. Overfitting is a sin
- 6. Honor thy data's first mile and last mile

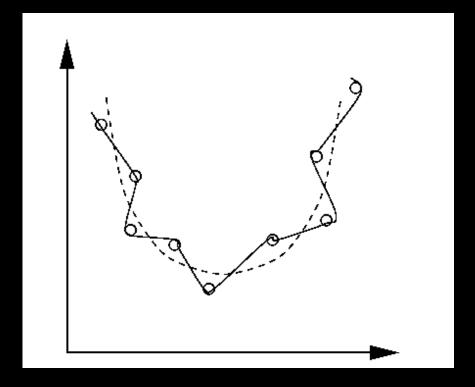
(a) The First Mile is the hardest : ...integrating ubiquitous heterogeneous data.
(b) The Last Mile is the hardest : ...extracting actionable intelligence. All of the features in the data histogram convey valuable (actionable) information (the long tail, outliers, multi-modal peaks, ...)



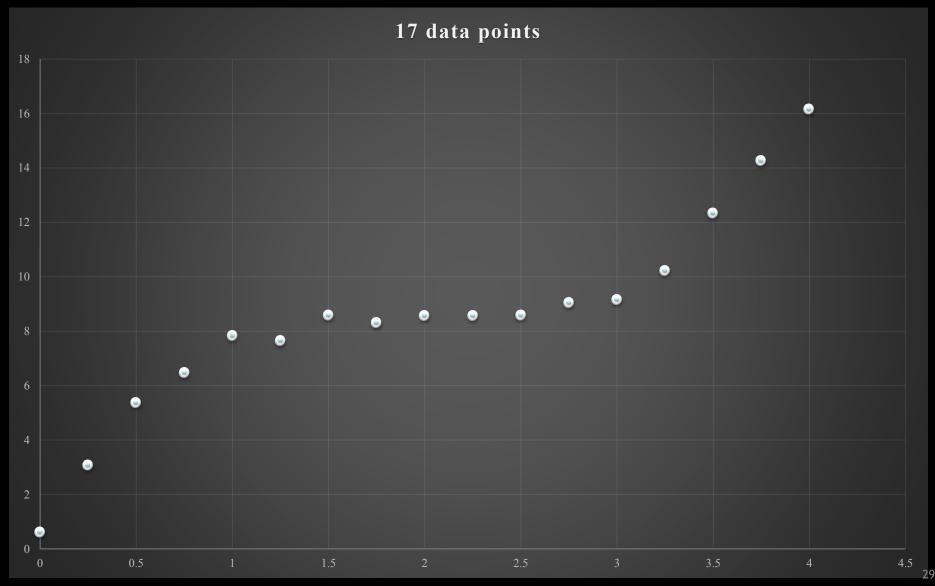
You can discover real insights in the long tail e.g., the median Twitter account has only 1 follower! http://radar.oreilly.com/2013/12/tweets-loud-and-quiet.html



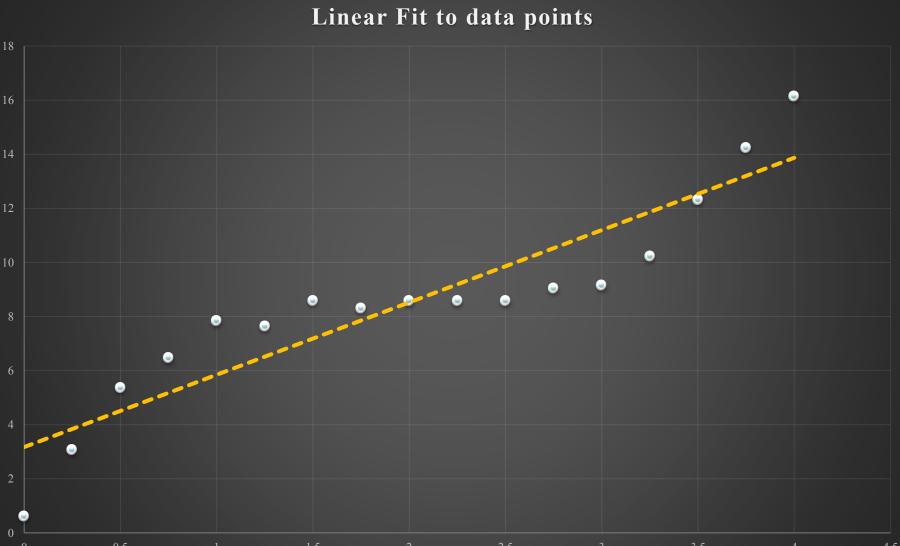
Overfitting is a sin



Data Literacy Lesson: Build a Predictive Model of your data points

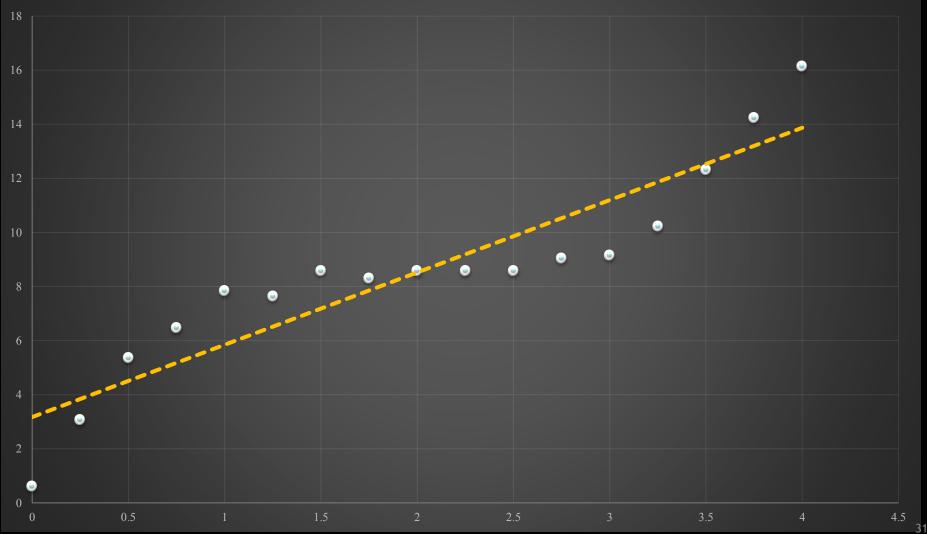


Data Literacy Lesson: 1st attempt – fit a line through the points



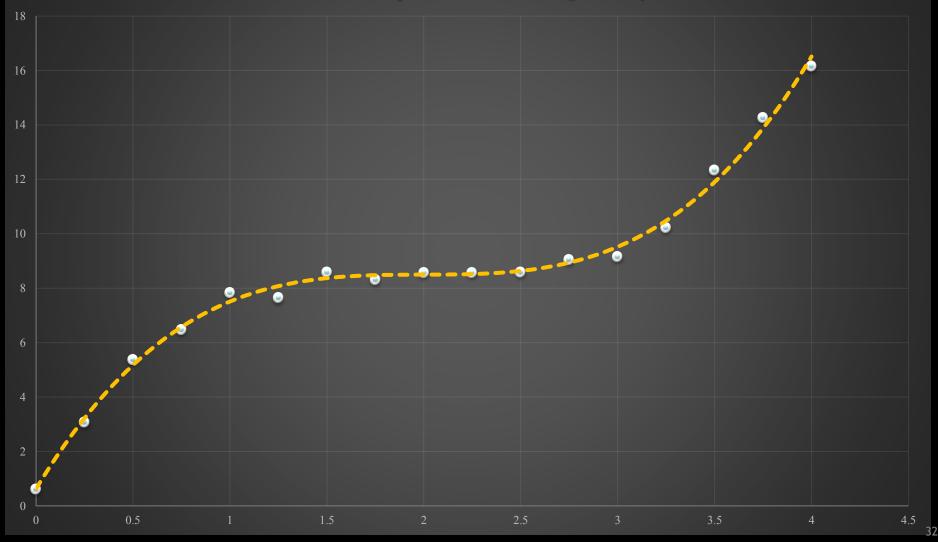
Data Literacy Lesson: 2nd attempt – use a quadratic polynomial

Parabolic fit to data points, something like: $y = x^2$



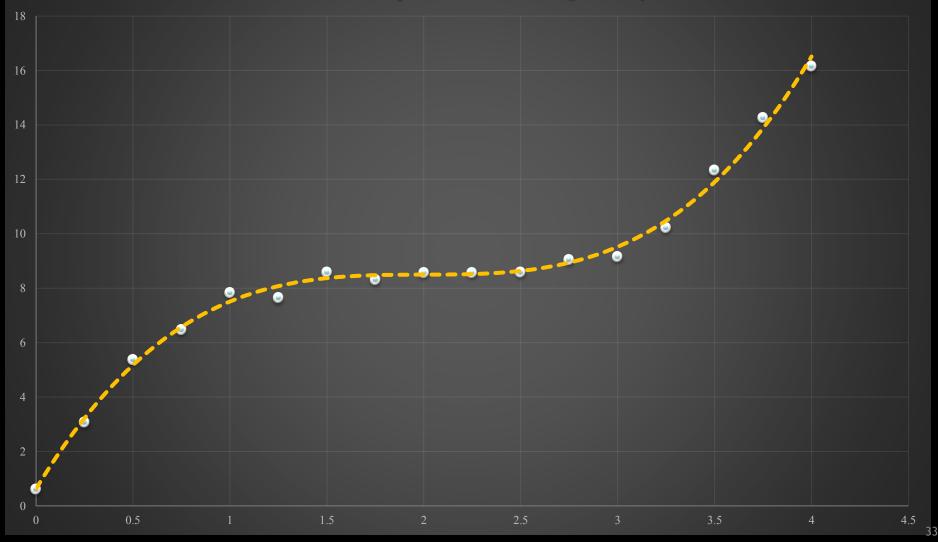
Data Literacy Lesson: 3rd attempt – use a cubic polynomial

Cubic fit to data points, something like: $y = x^3$

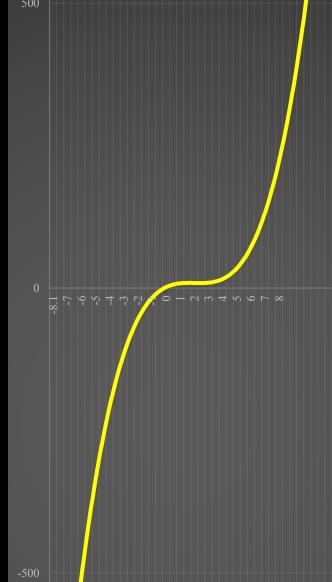


Data Literacy Lesson: ***** Is this really better?? *****

Cubic fit to data points, something like: $y = x^3$



Data Literacy Lesson: ***** Is this really better?? *****

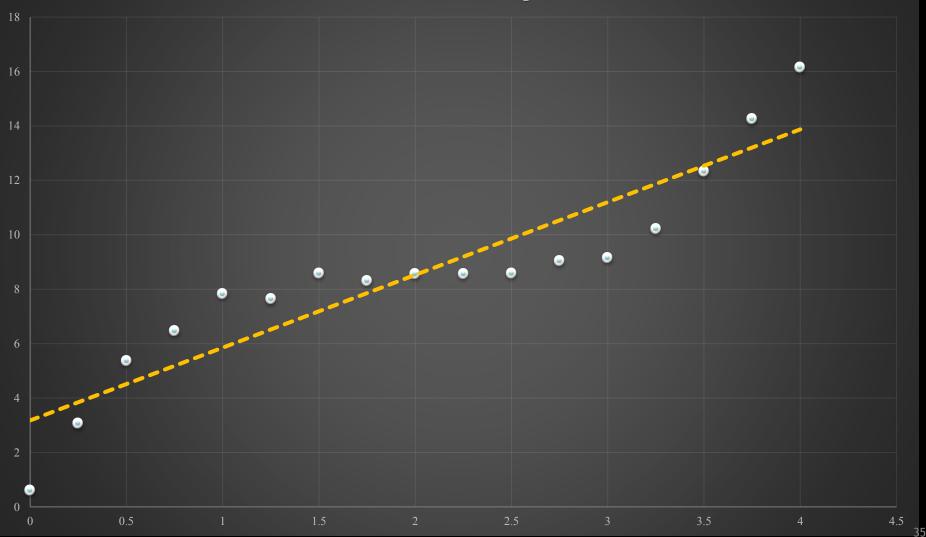


This is example of Overfitting the data!

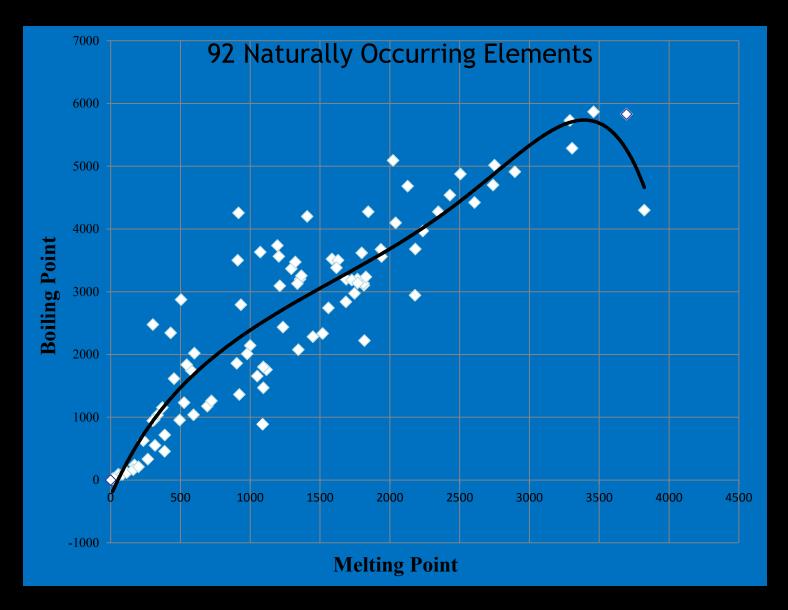
Data Science rule #1: Don't Overfit Your Model

Data Literacy Lesson 1st attempt was a Good Fit to data points

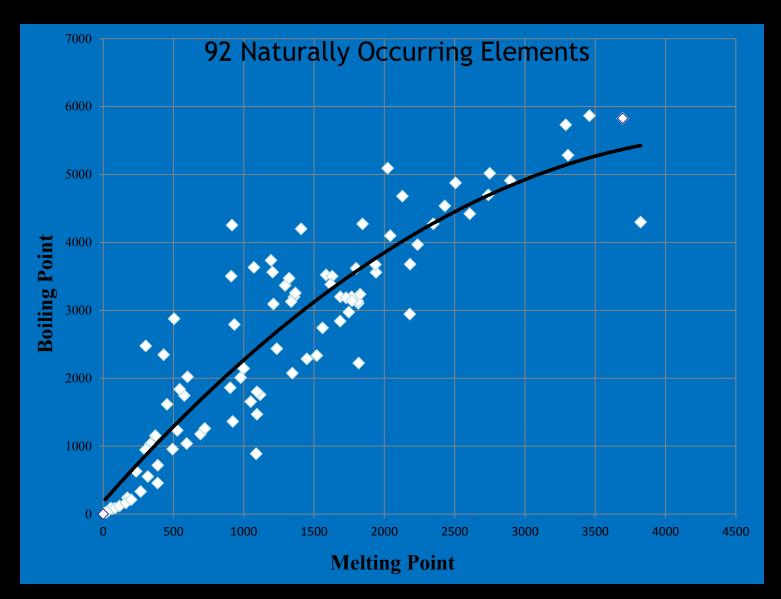
Linear Fit to data points



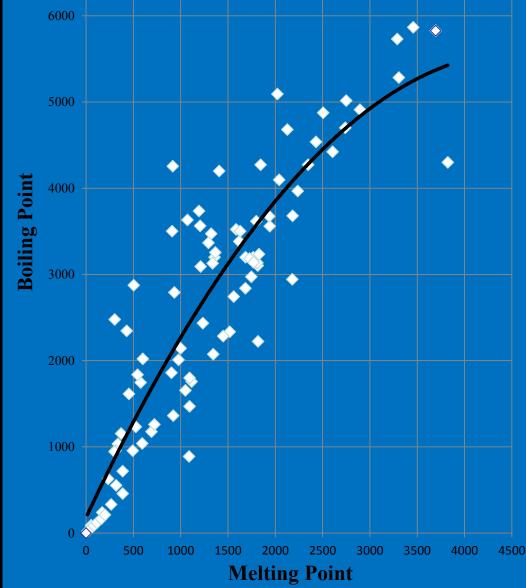
Trend Lines in big data sets: Descriptive Analytics! It is tempting to over-fit every wiggle in the data.



This is a better fit to the trend line... for use in Predictive Analytics!

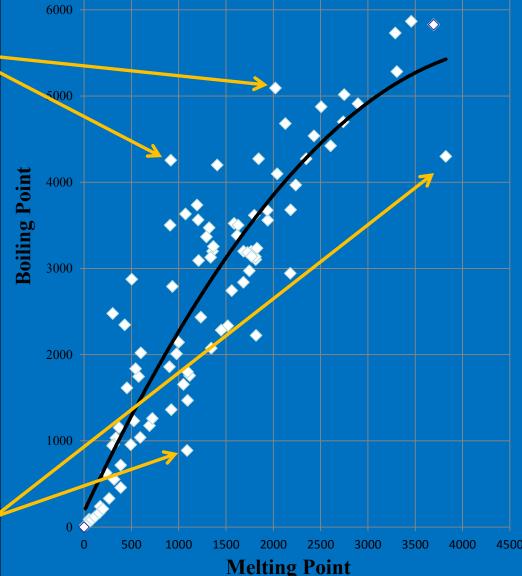


Trend Line



Trend Line and Outliers:

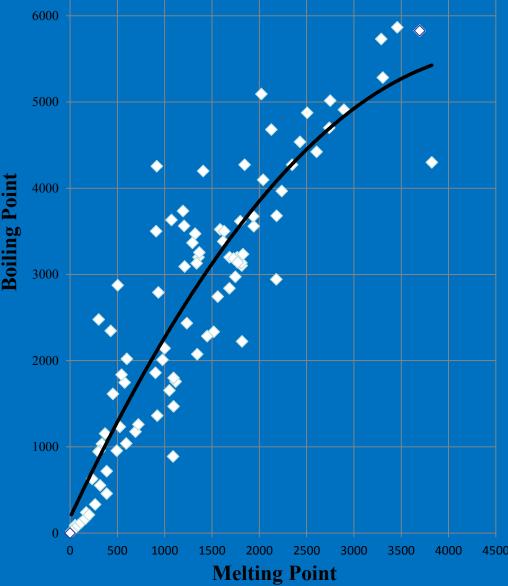
Boiling Points and Melting Points of the 92 Chemical Elements



Sometimes we are tempted to think that outliers are just noise or natural variance.

Trend Line and Outliers: where is the real discovery?

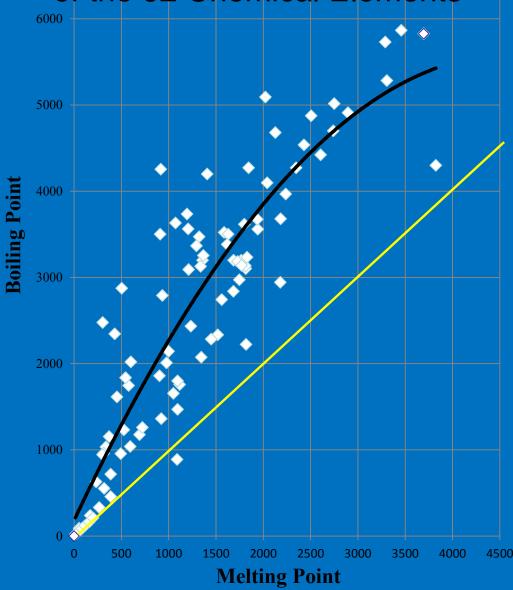
Sometimes we are tempted to think that outliers are just noise or natural variance.



Trend Line and Outliers: Add some context to the data!

...that diagonal line in the plot (where melting point = boiling point)

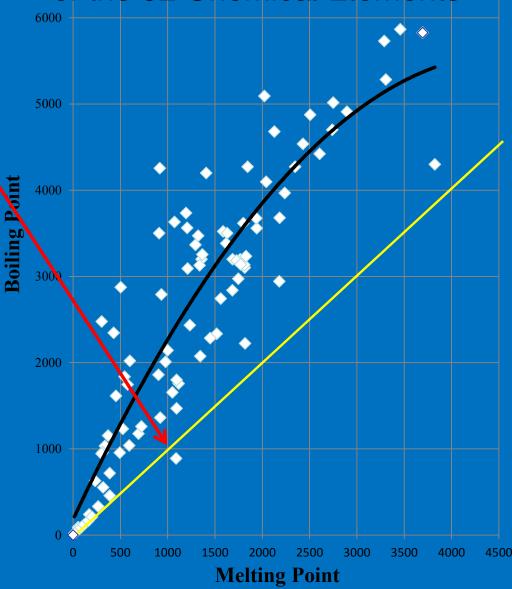
... this provides some context (related to your prior knowledge)!



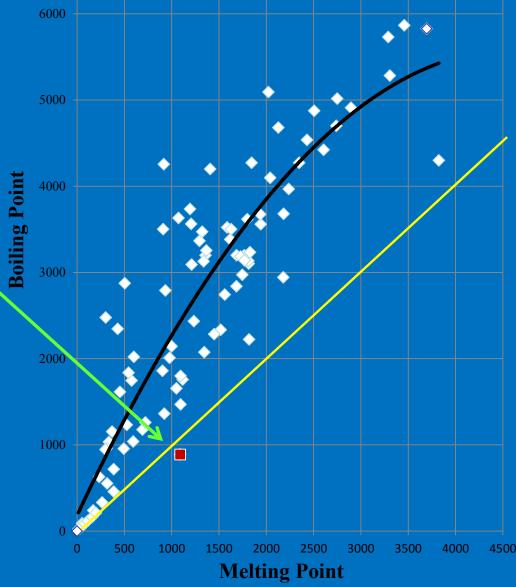
Trend Line and Outliers: What is that point below the line?

...that diagonal line in the plot (where melting point = boiling point)

... this provides some context (related to your prior knowledge)!

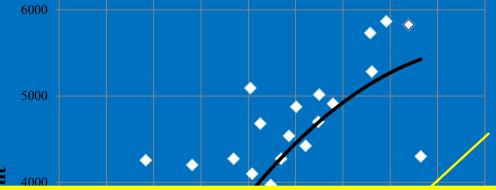


Trend Line and Outliers: there's the real discovery!



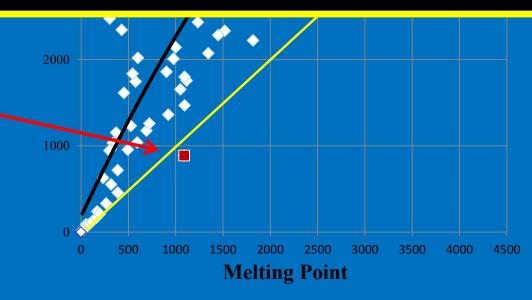
Trend Line and Outliers: there's the real discovery!

Boiling Points and Melting Points of the 92 Chemical Elements



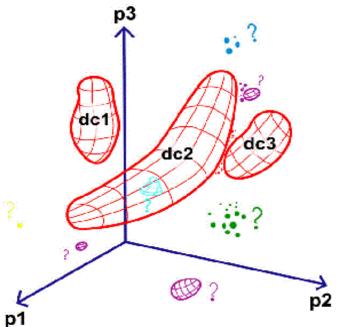
Surprise Discovery!

Melts @ 1089°K Boils @ 889°K -<u>Arsenic</u>!

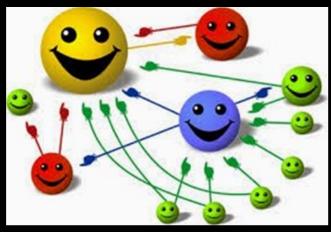


Data Literacy Matters: Using Data for Discovery in 4 flavors

Data Mapping and a Search for Outliers



Graphic provided by Professor S. G. Djorgovski, Caltech



 Class Discovery: Finding new classes of objects (population segments), events, and behaviors. This includes: learning the rules that constrain the class boundaries.

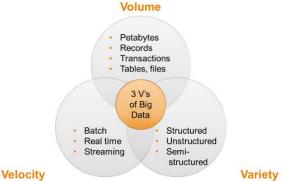
2) Correlation (Predictive and Prescriptive Power) Discovery: Finding patterns and dependencies, which reveal new governing principles or behavioral patterns (the "DNA").

3) Novelty (Surprise!) Discovery: Finding new, rare, one-in-a-[million / billion / trillion] objects and events.

4) Association (or Link) Discovery: Finding unusual (improbable) co-occurring associations.

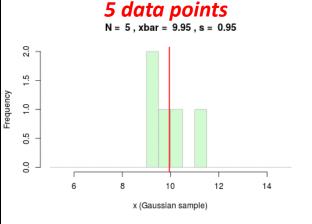
Big Data Challenges

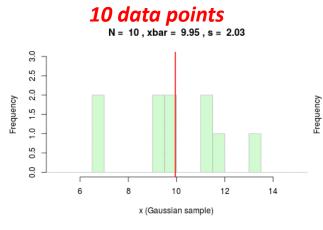
- The 3 V's of Big Data are not just hype they represent really big challenges:
 - **1. Volume**
 - Variety
 Velocity



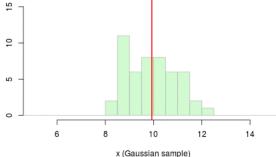
- But... Volume is not the problem! Storage is manageable.
- Data Science & Analytics (integrating and combining disparate data sources to achieve Data-to-Discovery, Datato-Decisions, and Data-to-Dividends) are hard...
- ... especially on complex (diverse, high-Variety) and fast-moving (real-time, high-Velocity) data!
- Enabling Advanced Analytics / Data Science capabilities is therefore the key to conquering these challenges.

Big Data Volume is great news... ... more data means less uncertainty, and more laser-focused insights & intelligence!



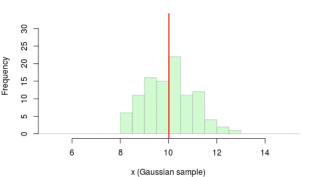






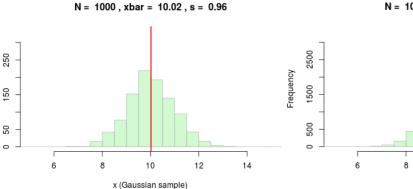
100 data points

N = 100 , xbar = 10.01 , s = 1



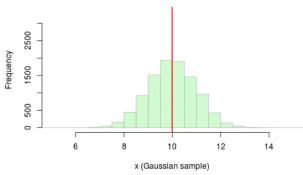
1000 data points

Frequency



10000 data points

N = 10000, xbar = 9.99, s = 1

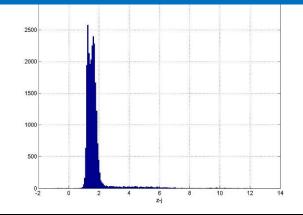


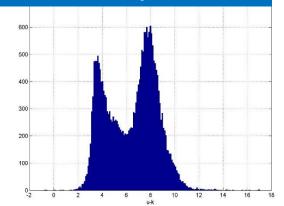
47

...but the greatest of V's is Variety

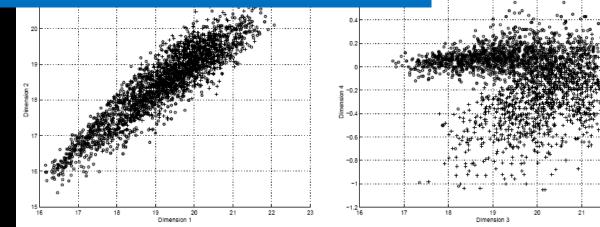
The discovery and separation of classes improves when a sufficient number of "correct" features are available for exploration:

(a) 2 classes are discovered and become separable:





(b) One trend line becomes 2 clusters:



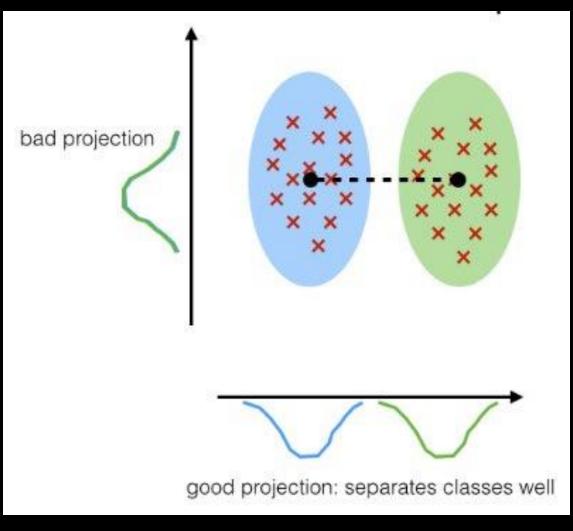
Clustering

(for class discovery, segmentation, personalization, SegOne marketing,...)

Exploiting the 3rd V of Big Data
(Data Exploration and Data Exploitation)
1. Volume
2. Velocity

3. Variety

Feature Selection and Projection



Feature Selection is important to disambiguate different classes. More importantly, **Class Discovery** depends on selecting the right features!

Feature Selection and Model Bias: choosing features in the dark



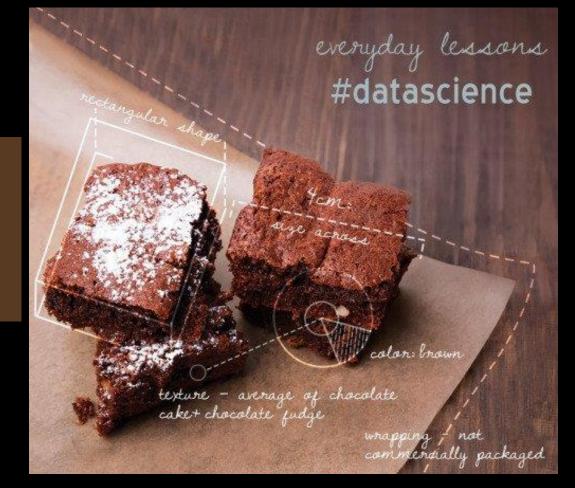
I picked out two socks from my sock drawer this morning!

It was still dark, but that shouldn't matter, right? After all, they are the same size ... THE SAME ?!?

The Era of Big Data represents the END OF DEMOGRAPHICS (*i.e.,* our models should no longer be based on and biased by a limited selection of attributes and features)

High-Variety Data enables better (and tastier) analytics models

Variety is the spice of discovery!



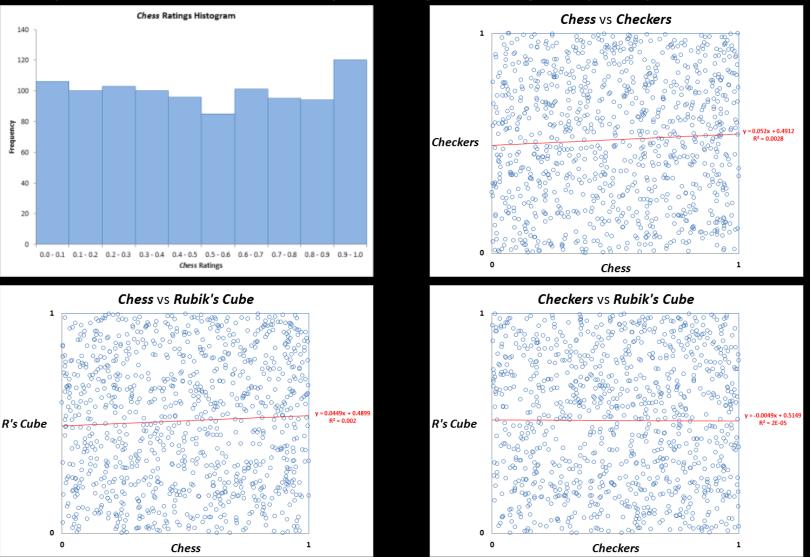
The Data Science of Feature-rich Chocolate Brownies http://www.datasciencebowl.com/data-science-of-chocolate-brownies/

A Statistical Data Puzzle



The Island of Games Puzzle: Can you find a pattern in the player ratings data?

http://www.datasciencecentral.com/profiles/blogs/island-of-games-puzzle-problem-statement

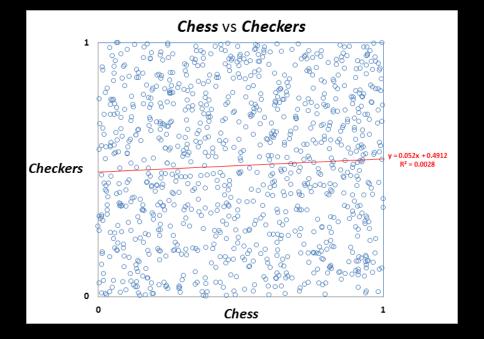


Solution to the Island of Games Puzzle

Island of Games

Color-coding the player ratings data distribution

Green and Red = High Rubik's Cube ranking (> 0.5); Blue and Yellow = Low Rubik's Cube ranking (< 0.5)

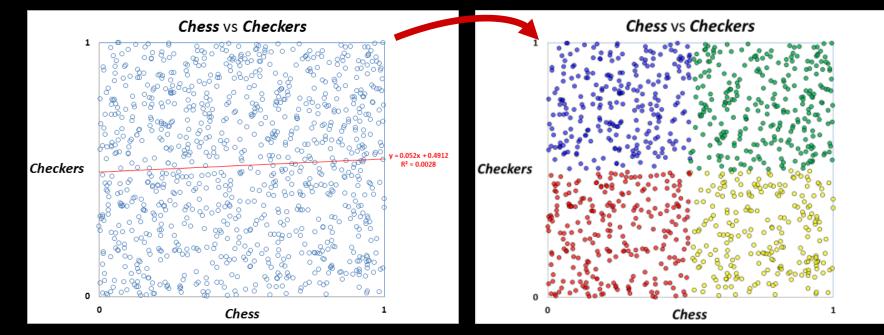


The intrinsic patterns in the player ratings data are not revealed in 2-D scatter plots or by using traditional statistical methods.

Island of Games

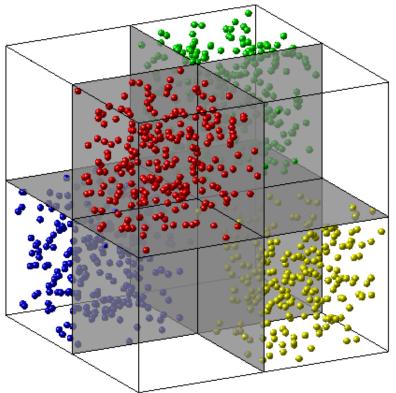
Color-coding the player ratings data distribution

Green and Red = High Rubik's Cube ranking (> 0.5); Blue and Yellow = Low Rubik's Cube ranking (< 0.5)



The intrinsic **patterns** in the player ratings data are not revealed in 2-D scatter plots or by using traditional statistical methods. However, exploration in the 3-D input parameter space (of player ratings for 3 games) reveals the actual player groupings...

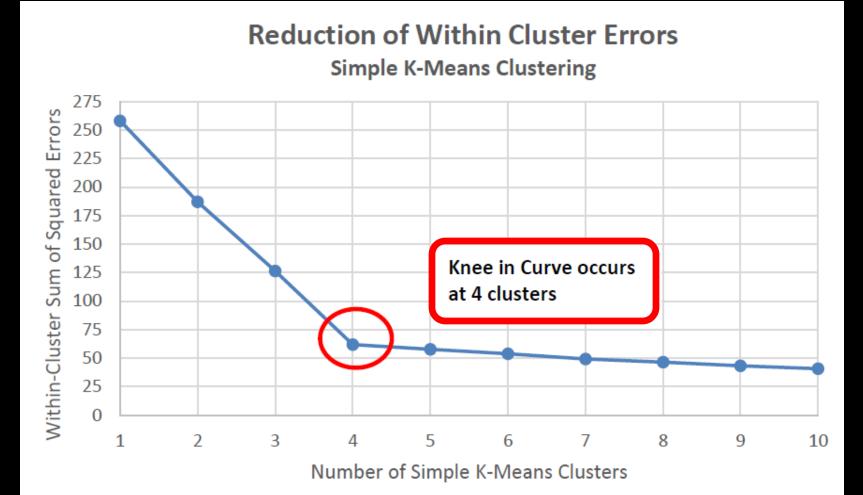
3-D view of the Player Ratings Data The true 3-D data distribution = 4 separable groups!



http://www.datasciencecentral.com/profiles/blogs/island-of-games-puzzle-problem-statement

Data Visualization Revelations

Solving the Island of Games Puzzle with a sequence of cluster models



Reference: Dr. Joseph Marr, GMU

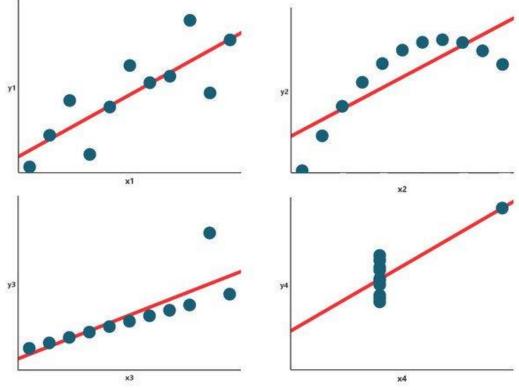
Further Data Visualization Revelations

Anscombe's Quartet demonstrates classic case where statistical parameters are truly unrevealing of the different distributions of these four data sets, but the data structure is finally revealed through data visualization!

Anscombe's Quartet

https://en.wikipedia.org/wiki/Anscombe%27s_quartet Identical Means, Variances and Correlation Coefficients

Linear regression line in each case: y = 3.00 + 0.500*x



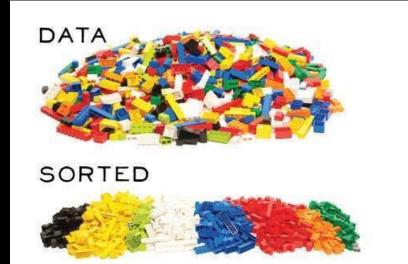
Mean of x in each case: 9, sample variance of x in each case: 11 Mean of y in each case: 7,50, sample variance of y each case: 4.122 to 4.127 Correlation coefficient between x and y in each case: 0.816 Linear regression line in each case: y= 3,00 + 0,500*x

DataStories @LindaRegber lindaregber.com

Start Young with Statistical and Data Literacy!

- Incorporate data & statistical literacies in every grade, in every type of course (not just math / STEM classes).
- Teach the 4 R's: Reading, wRiting, aRithmetic, and "R"
- Humans (even little ones) naturally characterize, sort, cluster, and classify by different observable attributes:





 Sign the Global Data Literacy petition: <u>http://oceansofdata.org/call-action-promote-data-literacy</u>

Data Literacy with Data Science will conquer Big Data's big challenges:

a) Volumeb) Varietyc) Volume



The 3 new V's of Big Data will become: 1) Veni (I came) 2) Vidi (I saw) 3) Vici (I conquered) Let's conquer the world's problems together with Data & Statistical Literacy!

Booz | Allen | Hamilton

NextGen Analytics and Data Science

Harnessing the Power of Data

Ten Signs of Data Science Maturity



Peter Guerra & Kirk Borne

O'REILLY

Tips for Building a

Data Science Capability

Some Hard-Won Best Practices and (Surprising) Lessons Learned

Booz | Allen | Hamilton

1

LISTEN @KirkDBorne @DataSci4Good @BoozAllen

READ

www.boozallen.com/datascience

- **The Field Guide to Data Science**
- Building a Data Science Capability
- 10 Signs of Data Science Maturity

PARTICIPATE

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THE FIELD GUIDE TO DATA SCIENCE

Passion. Curiosity. Purpose.

