

Data Science and Technology Entrepreneurship

Course Recap

Week 14 Sameer Maskey

Friday Open Office hours + Co-work with teams

- Come and work together
- Ask questions
- Dates :
 - April 19 4:30 6:30 pm
 - April 26 4:30 6:30 pm
 - May 3 4:30 6:30 pm
- Location
 - CS conference room

- Final Pitch/Demo Guidelines
 - 4 min long
 - Judges will provide feedback in judging time
 - I2 concurrent demos

2 laptops

- I for slides
- I for demo

Final Pitch/Demo Guidelines

- Executive Summary
- I-3 page executive summary
- Make 15+ copies that judges can take home

Executive Summary

Please write a 1-3 page Executive Summary. You can use parts of your Final Report and Business Canvas for the executive summary. You can use either of the following template: <u>http://www.entrepreneur.com/formnet/form/</u> <u>653</u> OR <u>http://www.entrepreneur.com/formnet/form/459</u>

- Final DSTE Survey
 - I0 Questions
 - Provide feedback so that I can improve the course for Spring 2014
 - <u>https://www.surveymonkey.com/s/DSTE2013finalsurvey</u>

- Assignment 4 is out
- Due May 5th Sunday @6pm
- ▶ 4 parts
 - Part I Final Report
 - I have written down 16 points or so that should guide you on how to write the Final Report
 - (Important) Please provide the final business model canvas.
 - Write a paragraph on individual role and contribution of each team member to the business.

Assignment 4

Part II - Final Technical Report

- System Architecture Diagram
- Overall choices you made

Part III - MVP

- Last iteration of your MVP
- Please write about the changes you have made from the last MVP

Part IV - MVP Field Test

Similar to last homeworks - customer validation of your MVP with at least 5 customers

Final Presentation

- Date : May 7th (Tuesday)
 - Yes it's on **Tuesday**
 - First day of reading week
- Location Uris 142
- Schedule
 - ▶ 12:45 1:00 pm Registration
 - I:00 I:15 Overview of the class and projects
 - I:15 I:30 Mentors/Advisors Introductions
 - I:30 2:30 Student Presentation (4 min each)
 - > 2:30 3:15 Coffee and Sandwiches
 - 3:15 4:15 12 concurrent Student Demos (judges score the demos)
 - 4:15 4:30 Winners announcement



- Course Recap
- Survey
- Analyzing team effort

Team

- Imagine a company is willing to buy you for 100,000\$
- Split the money and give it to individual members
- Each person should send me what they believe is the right split

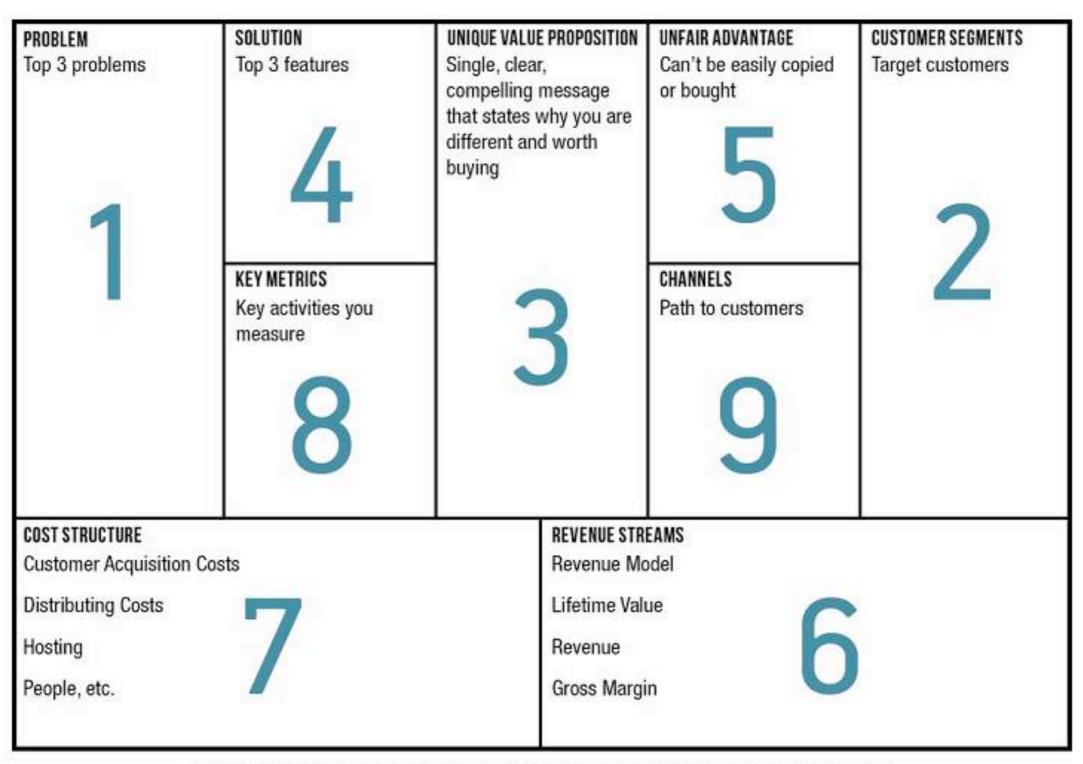
Customer/Market Risk vs Invention Risk

-	Market Risk			→ ←	Both			Invention Risk			
Web 2.0	Enterprise Software	Enterprise Hardware	Comm Hardware	Comm Software	Consumer Electronics	Game Software	Semicon	Electronic Design Automation	Cleantech	Med Dev / Health Care	Life Science Biotech

Source : Steve Blank

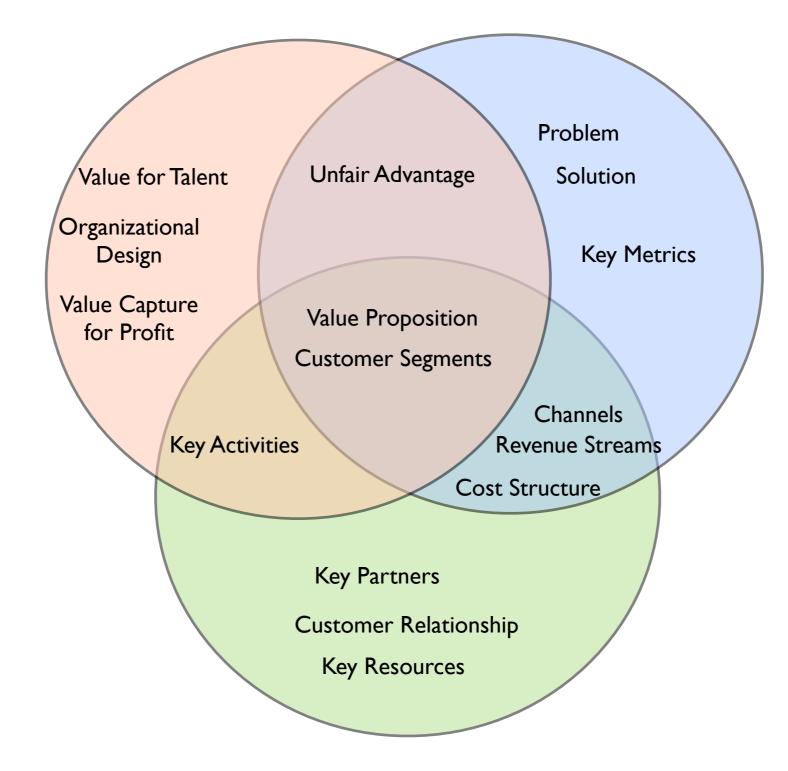
Where does your startup fall?

Lean Canvas [Maurya, A]

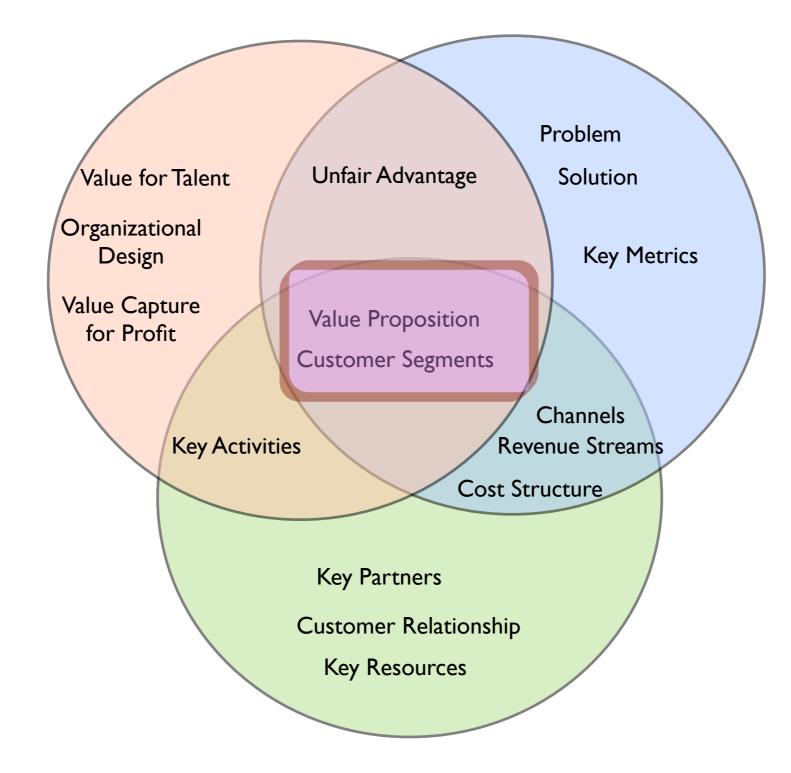


Lean Canvas is adapted from The Business Model Canvas (http://www.businessmodelgeneration.com) and is licensed under the Creative Commons Attribution-Share Alike 3.0 Un-ported License.

Common Concept Across Frameworks

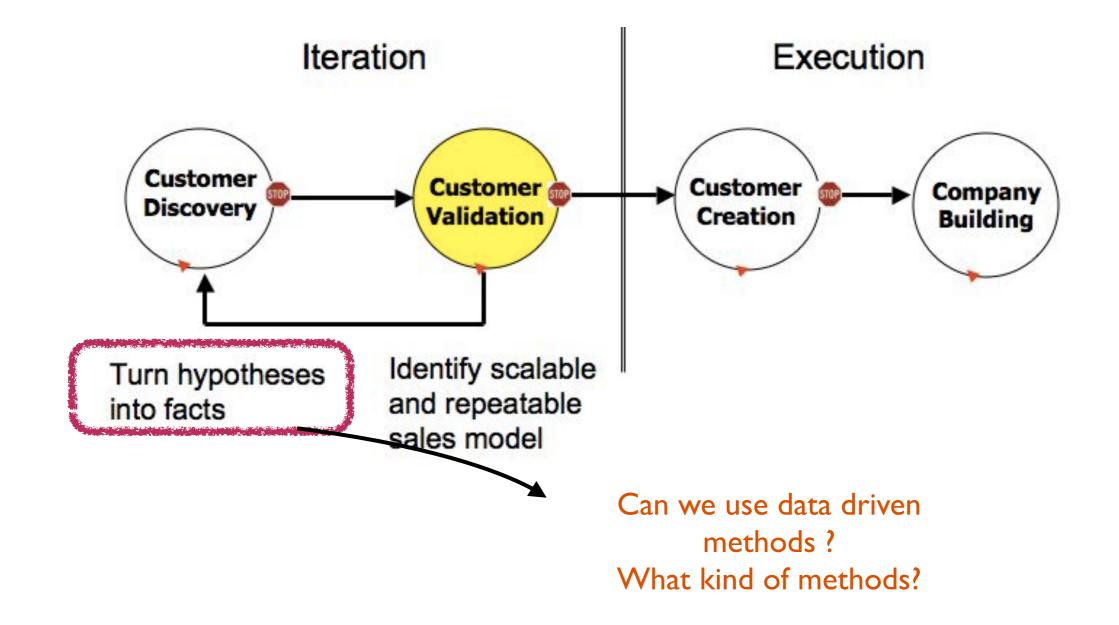


Common Concept Across Frameworks



Customer Discovery Process

Source : Startup Owner's Manual - Steve Blank and Bob Dorf



Data Driven Decision and Startups Hypothesis Design Insight **Experiments** Steve Blank's Customer **Development Insight** Test Some startups are cycle [Source : The already data driven Startup Owner's Manual]

Value Proposition & Customer Segments

This is common theme across all business model frameworks

2 important questions to ask yourself

- Who are your customers?
- What is the value proposition for them?

Customer Discovery with MVP

Phase I : Set of Hypotheses about your business (Problem?, Solution? Value Proposition?)

Phase 2 : Set of Hypotheses about your business (Test your hypotheses by talking to customers)

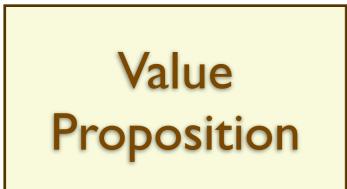
Phase 3 : Build MVP and test MVP with customers (Does your MVP solves the problem customer want?)

Phase 4 : Analyze results of your Phase 3 (Ready to signup paying customers?)

Multiple MVPs

- Multiple MVPs can be used to test competing hypotheses
- Example :
 - MVP with pay per use model
 - MVP with pay per month model
- If it is not difficult to build multiple MVPs then build them and test them with customers

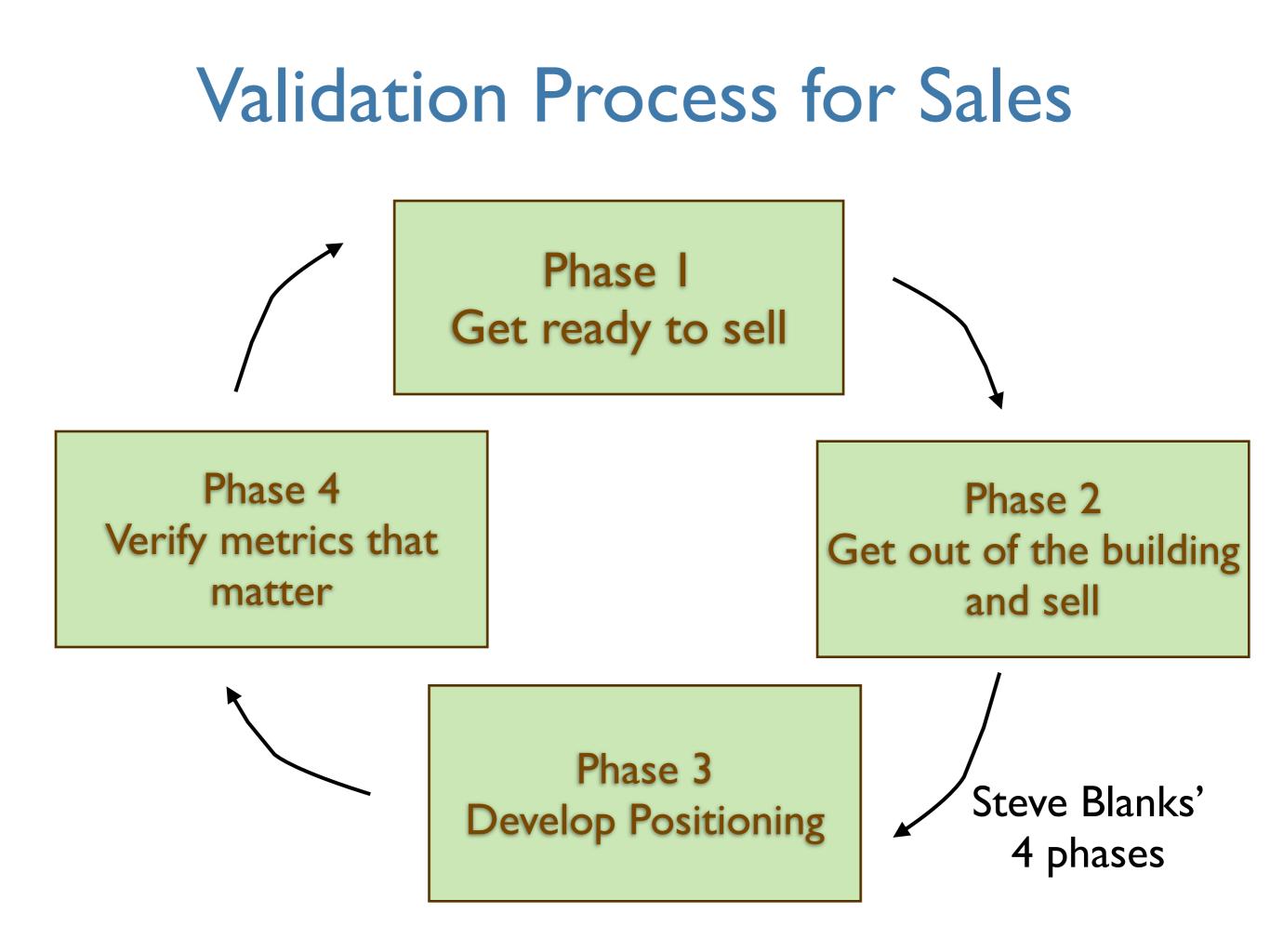
MVP and Sales













Is the business scalable?

- Sales roadmap
 - Is it repeatable?
 - Is it scalable?

Sales funnel predictable?

What are your validation checkpoints?

▶ B2B

- 3 meetings required for sale
- I0% responds to email
- 50% of respondent gives us meeting
- 50% of meeting result in contract signing meeting
- 50% of them use prototype
- 50% of them start paying
- Number of orders per month?
- Buys extra features

What are your validation checkpoints?

B2C

- 40% response rate to marketing email
- 20% sign up
- 30% are repeat visitors?
- > 20% of customer will pay for the service?
- I 0% will pay for more features?

What kind of Data Science Methods can you Use?

- As you run experiments for validation of your hypotheses you start generating data points
- You also generate a lot of data points from user engagements, sales engagements, etc
- How can you use some these data points?

Data to Insights

Data to Insight

Data to Scores

example: predict the likelihood of selling a watch

Data to Classes

example: predict what watches to buy from wholesaler

Data to Clusters

example: cluster customers based on their preferences

Machine Learning and Business

- Methods to analyze data that are all useful in decision making for businesses in general
- Data to Scores
- Data to Classes
 - Discriminative Methods
 - Generative Methods
- Data to Clusters

Use of Data Science Methods in Business

LinkedIn

- People You May Knows
- Identified number of connections it takes for a long term engagement
- Netflix
 - Signup process
 - Encourage to add movies to your queue
 - Once you add certain number of movies likelihood of you being a long term customer goes up

Use of Data Science Methods in Business

Zynga

- Monitors their users constantly
- Analyze how users interact with games to find out what makes a game successful
- Financial Services
 - Fraud detection
- OkCupid
 - Marketing Analytics with viral blogs
 - Facial attitude and new contacts blog

Fechnology Startups can generate a lot of data

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- For example a web startup with 500K users can generate a lot of data every user action is stored
 - Visits
 - Click through rates
 - Search logs
 - User generated content
 - Time spent on individual pages
 - Mouse movement behavior
 - Many more individual data points

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- Let's do another hypothetical example
- Using data science for a web startup that sells products online
- Want to increase the click through rate on related items?
- In other words, want to build a very simple minded recommendation engine

Example : User Data and Engagement

Assume you are running a shopping site and you want to produce top 5 items to recommend like Amazon

Inspired by Your Shopping Trends





RF Wireless Laser Pointer with Page... Generic

Satechi SP400 Smart-Pointer 2.4Ghz RF... (359) \$49.99 \$34.99



August LP103R Red Laser Presentation... August LP103R Red Laser (8) \$16.99 \$9.95



Logitech Wireless Presenter R400 (216) \$49.99 \$36.63

View your shopping cart

Data you have

Click through rates and data item

User Data History

Data from User's history

Product Clicks



Stauer Compendium Hybrid Watch

\$29.00 + No Shipping Info STAUER



5

4

Michael Kors Watches Bradshaw Gold - Michael \$250.00

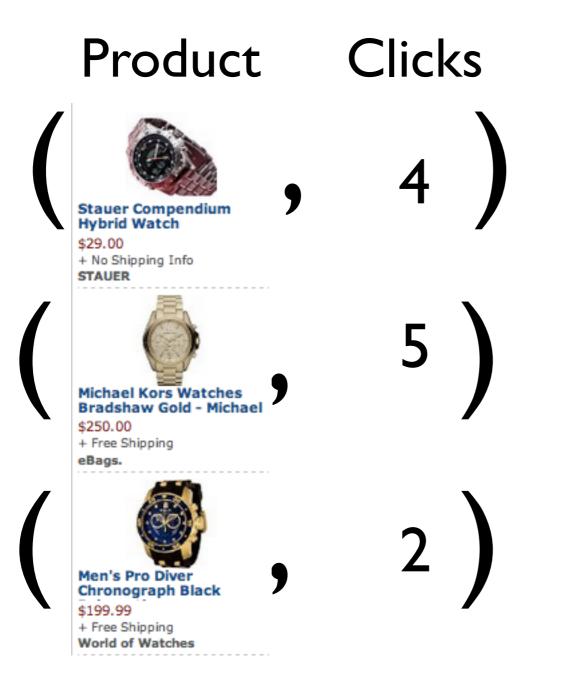
+ Free Shipping eBags.



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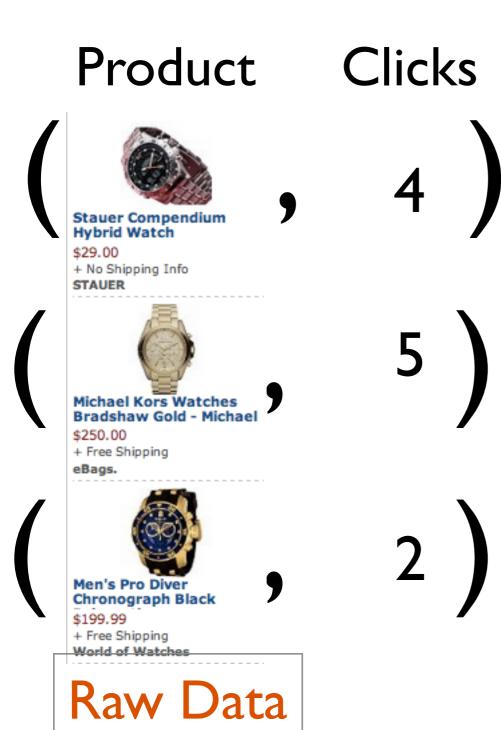
Men's Pro Diver Chronograph Black \$199.99 + Free Shipping World of Watches 2

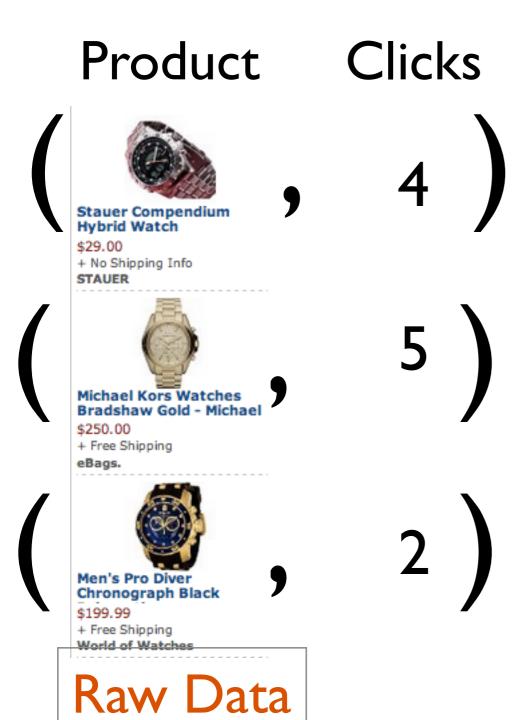
User Data History



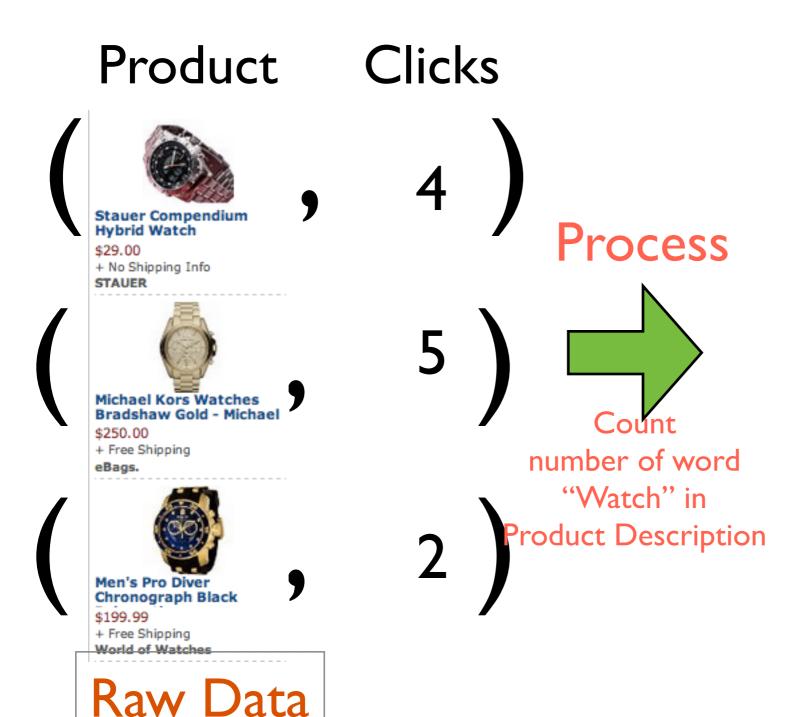
Can you use this data to build a simple model that can predict the number of clicks for a new product?

Imagine you have such data for millions of users





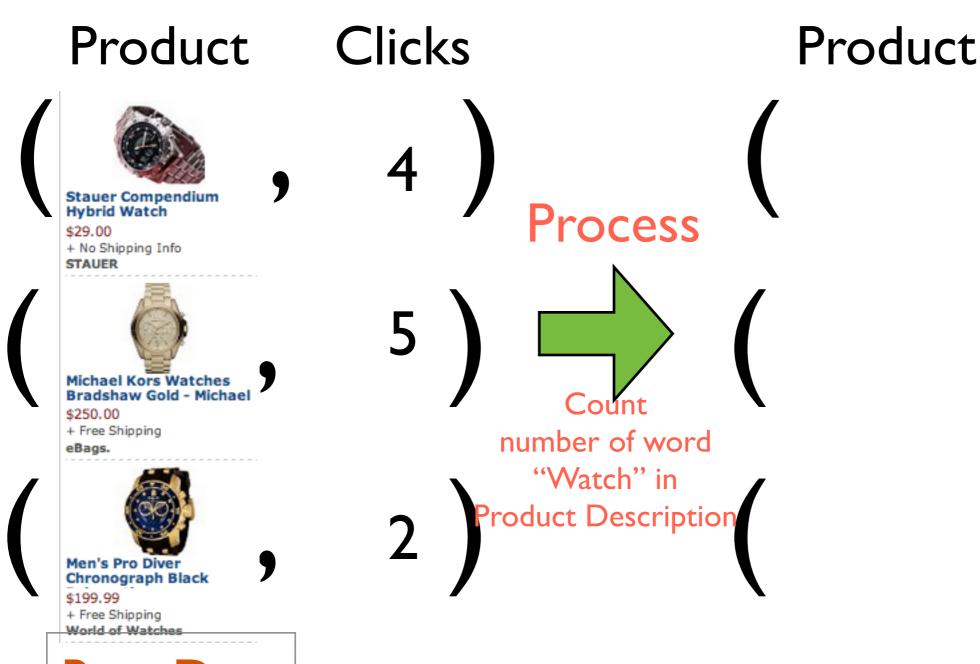


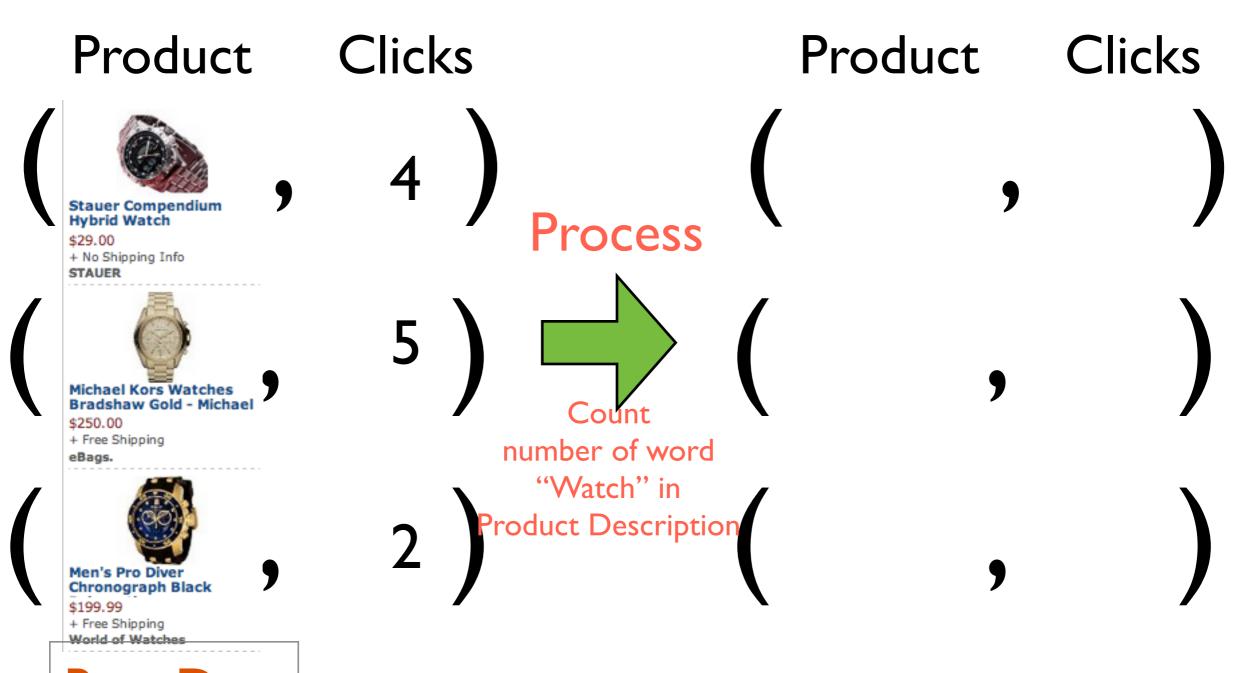


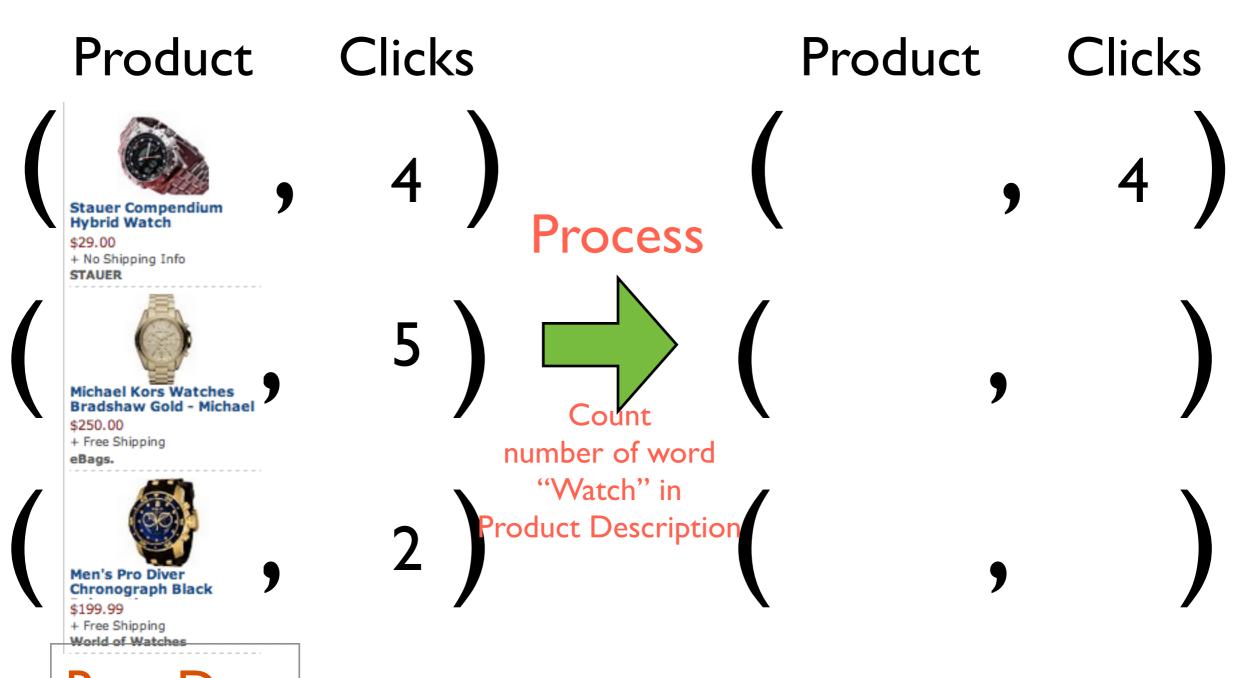


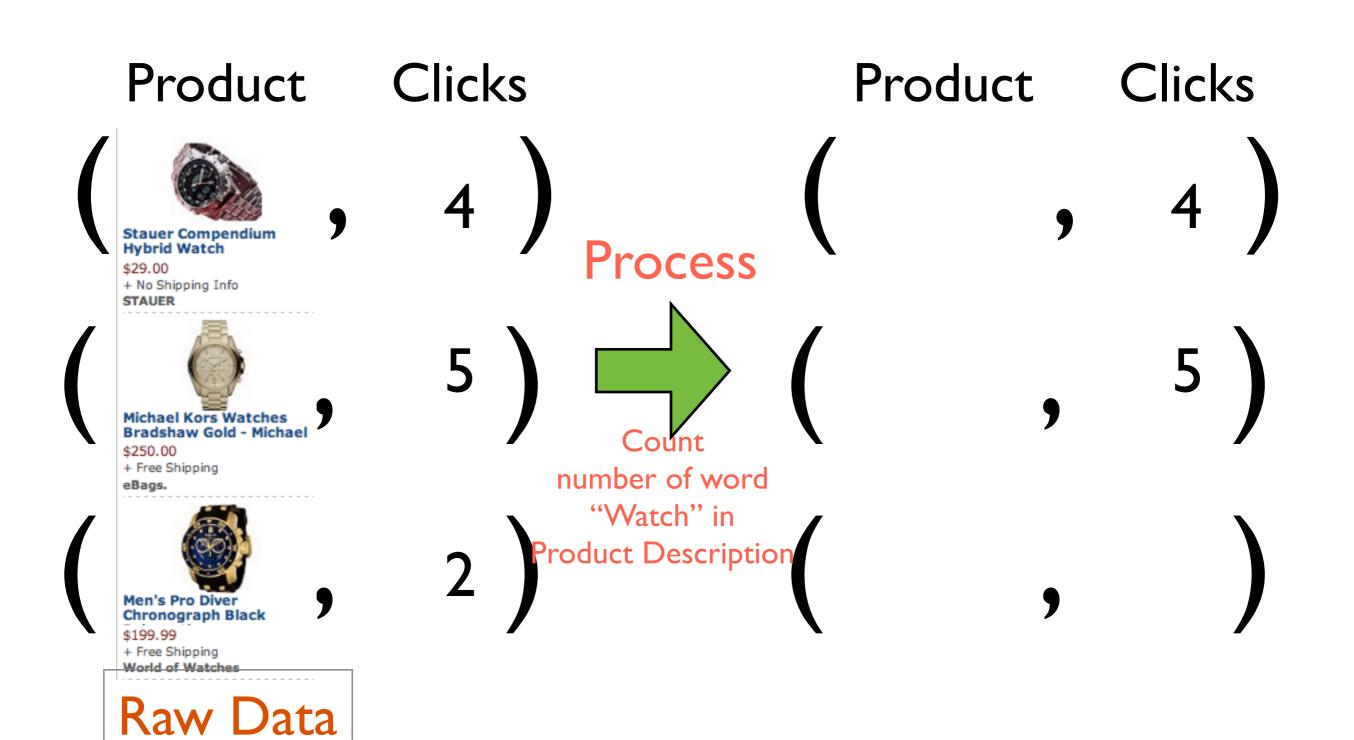


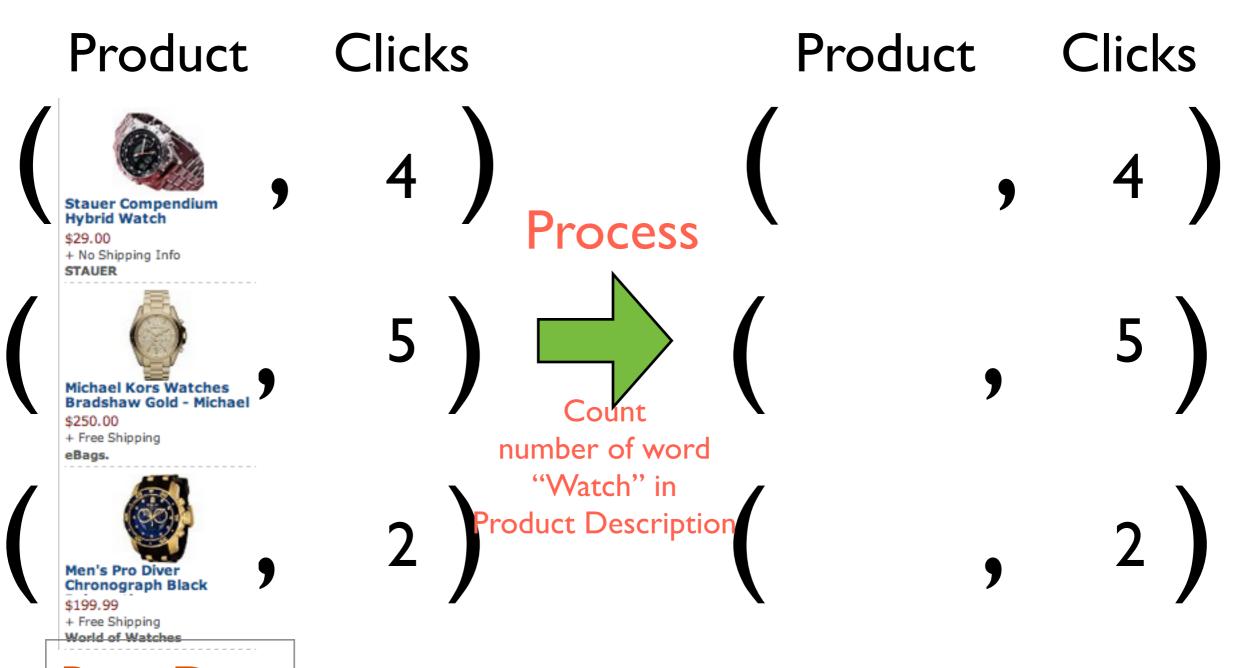


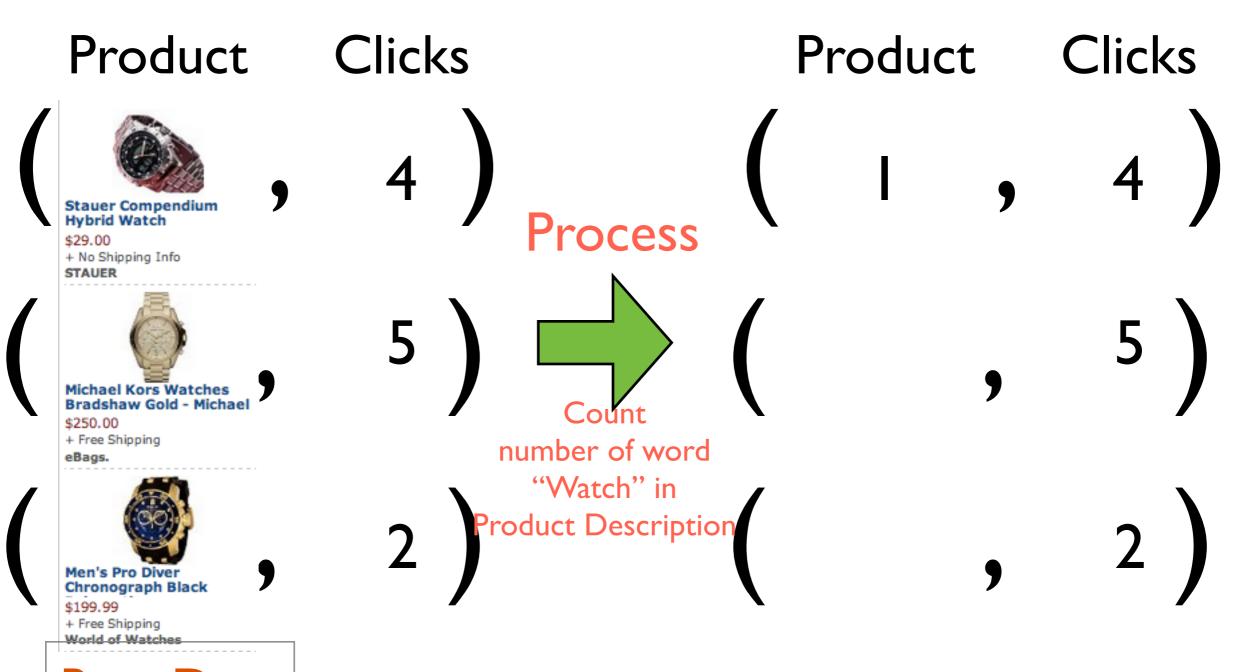


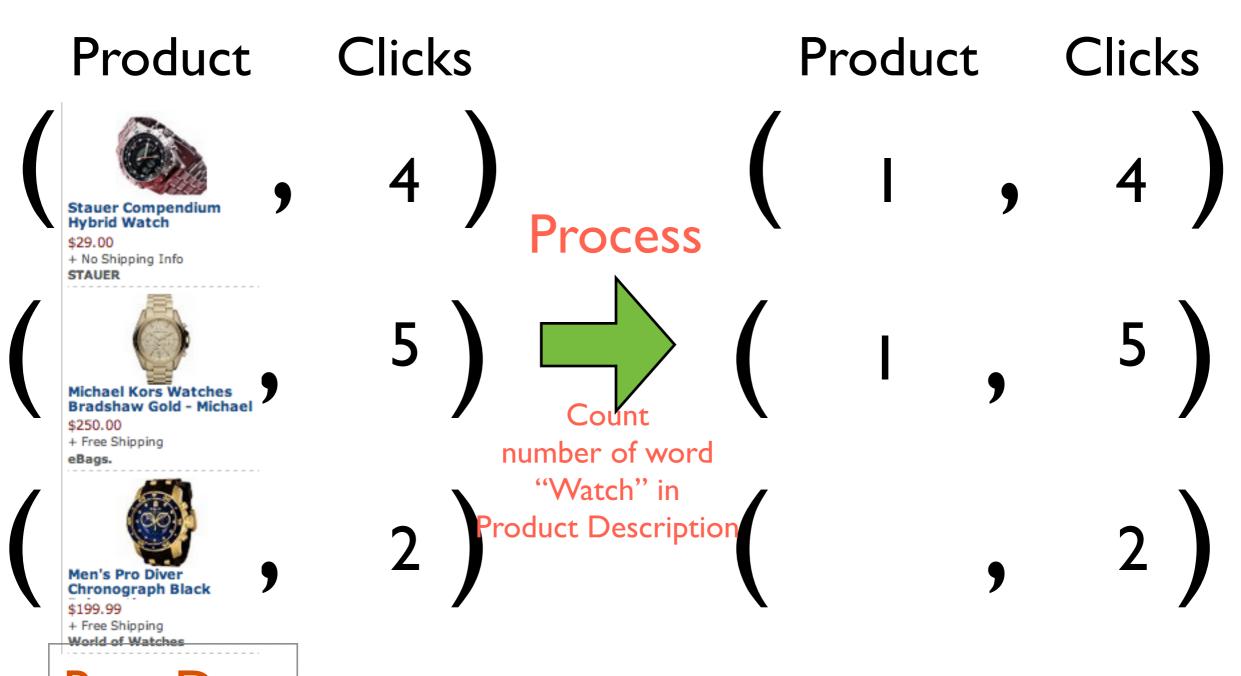


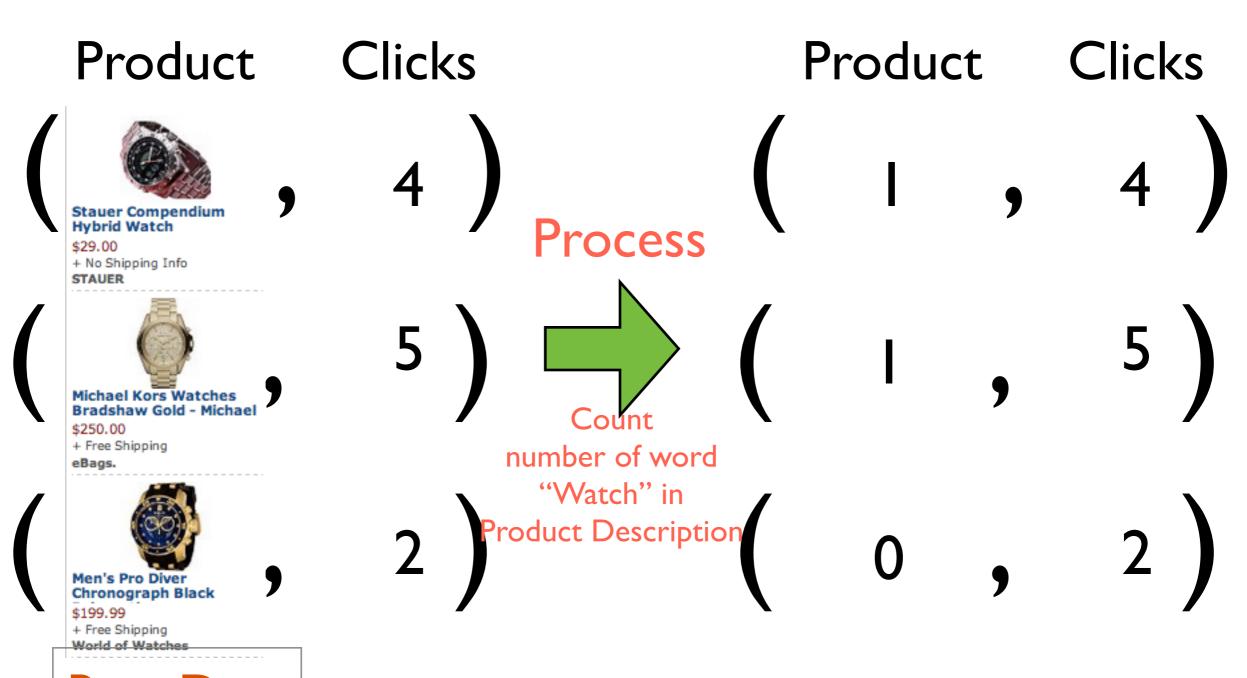


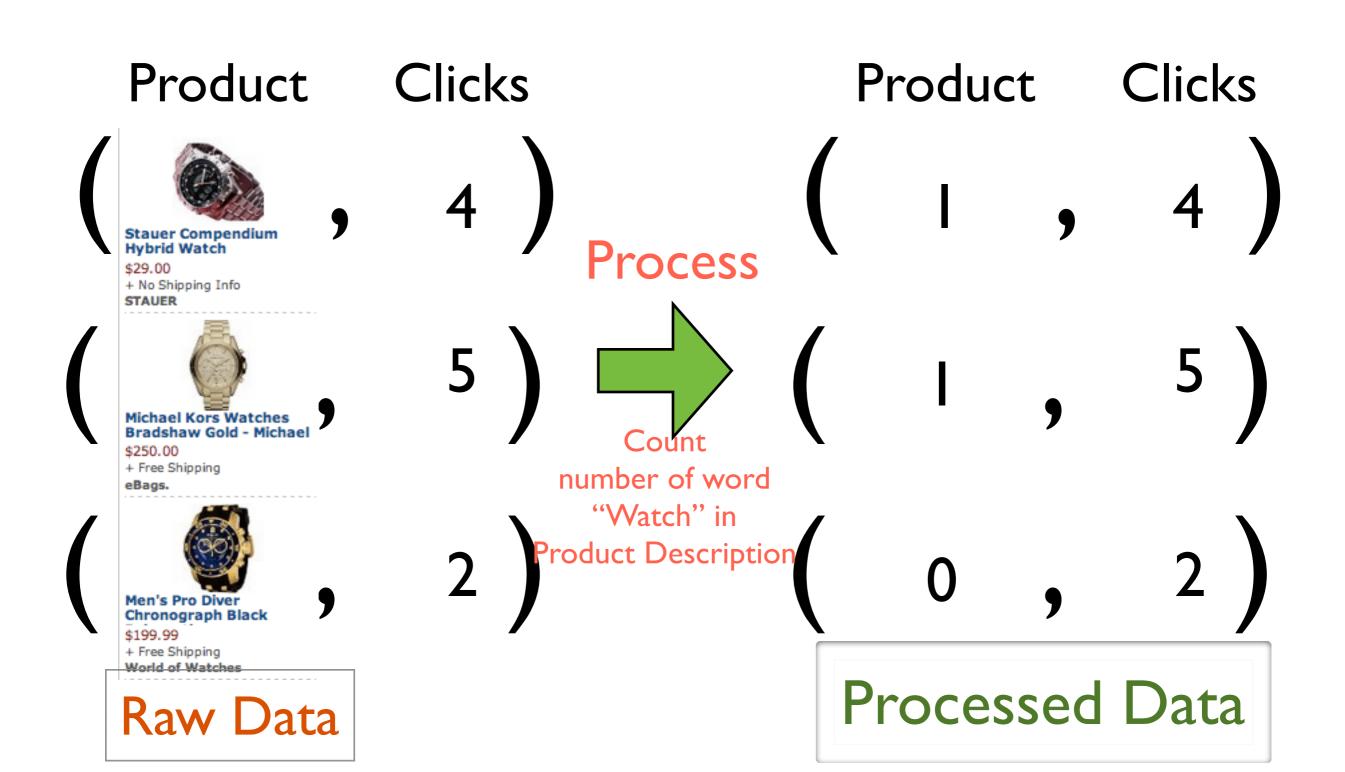


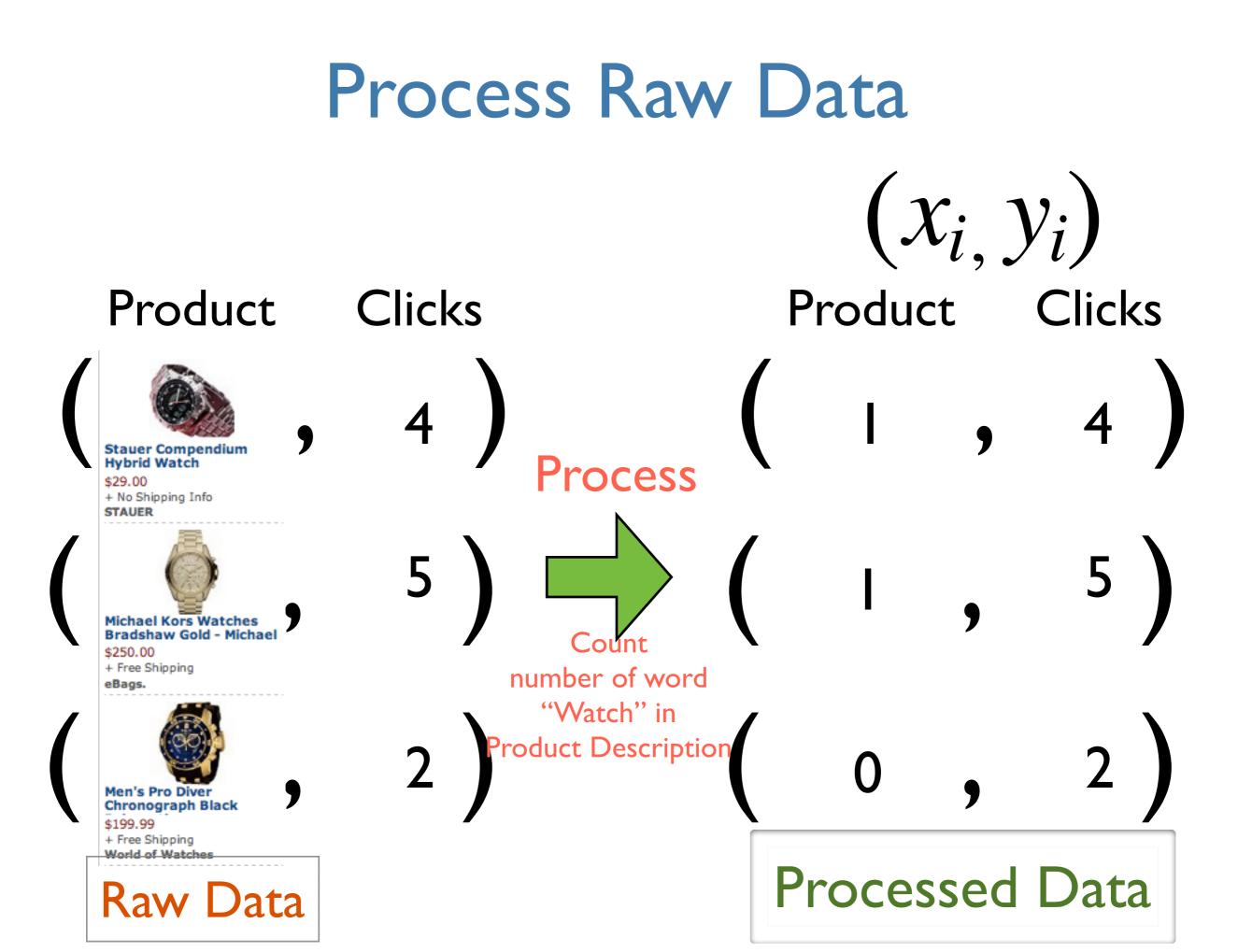




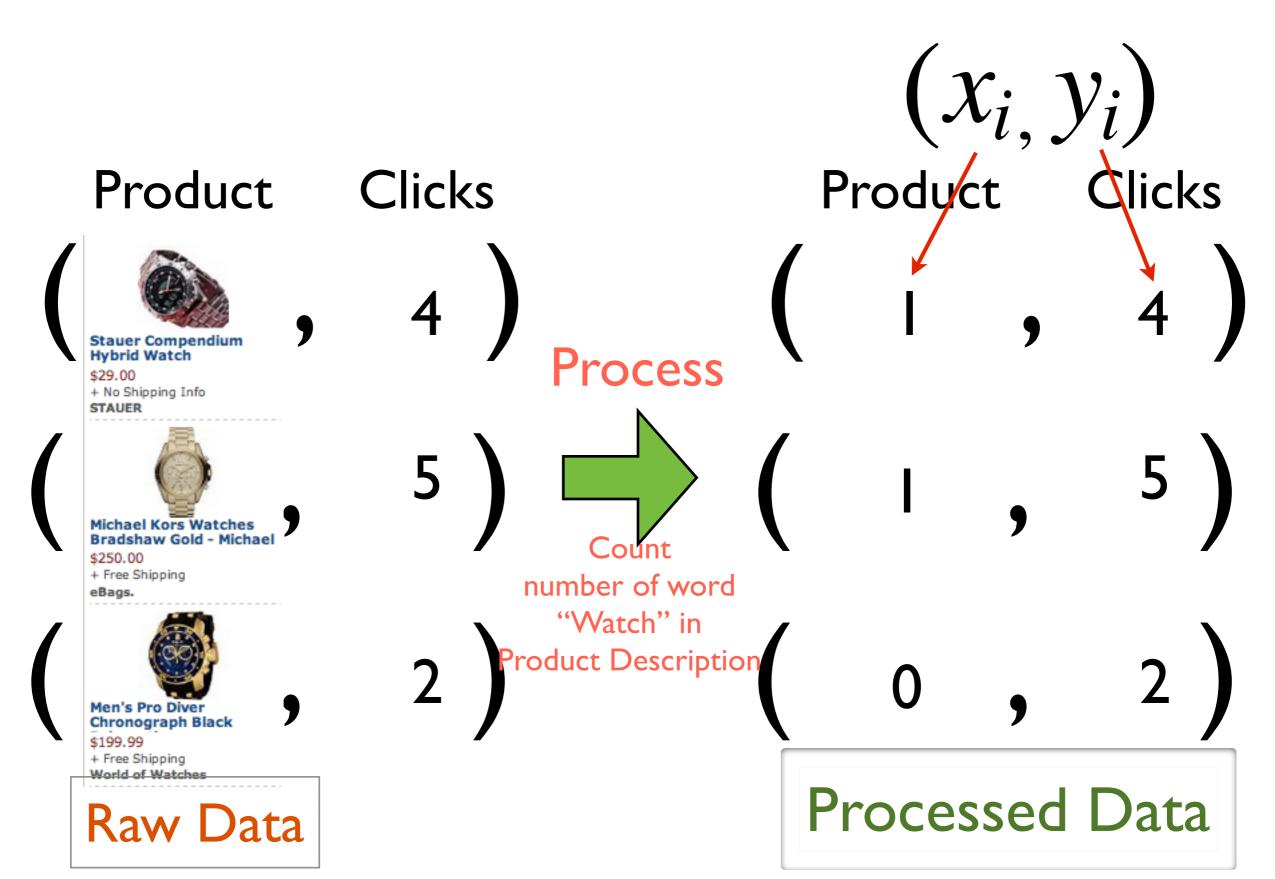


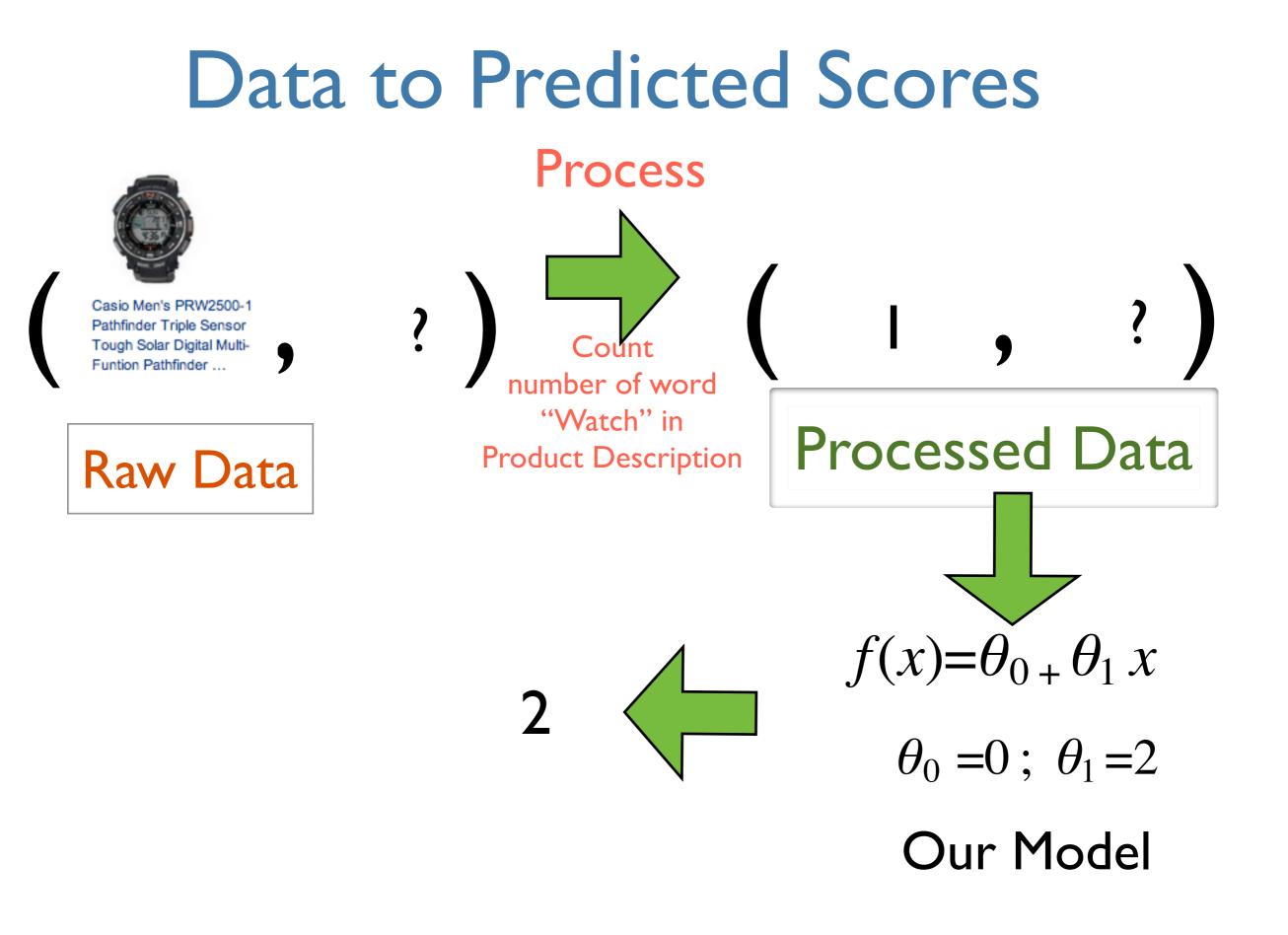




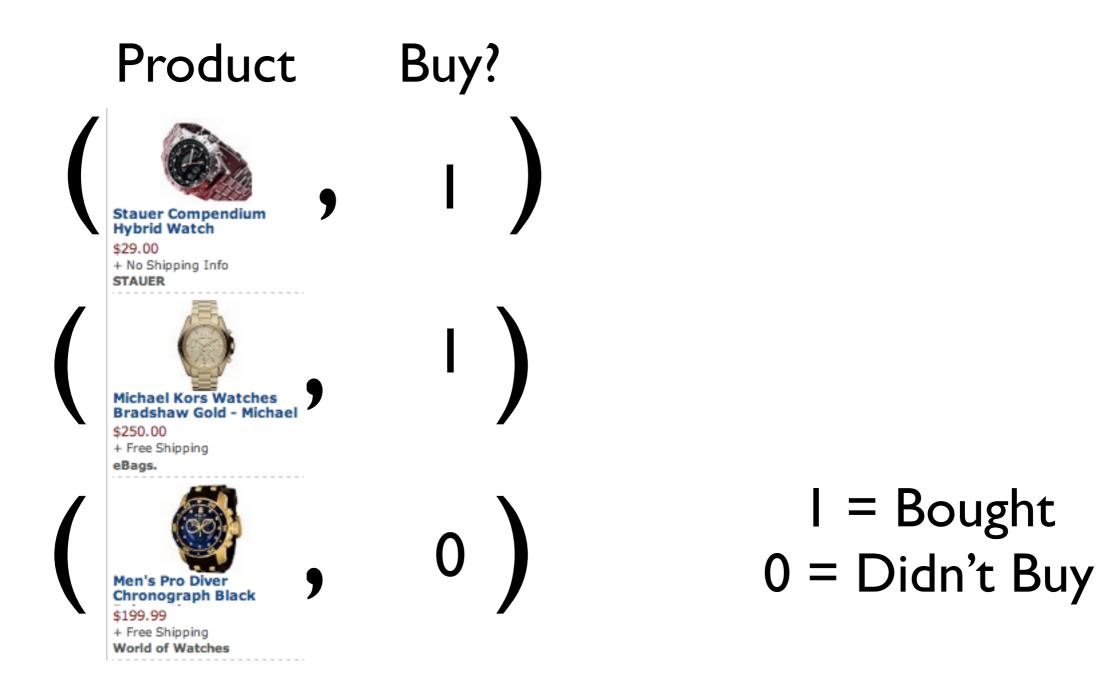


Process Raw Data (x_i, y_i) Product Clicks Product Clicks 4 Stauer Compendium **Process Hybrid Watch** \$29.00 + No Shipping Info STAUER 5 Michael Kors Watches Bradshaw Gold - Michael Count \$250.00 + Free Shipping number of word eBags. "Watch" in Product Description 2 ()Men's Pro Diver **Chronograph Black** \$199.99 + Free Shipping World of Watches **Processed** Data Raw Data

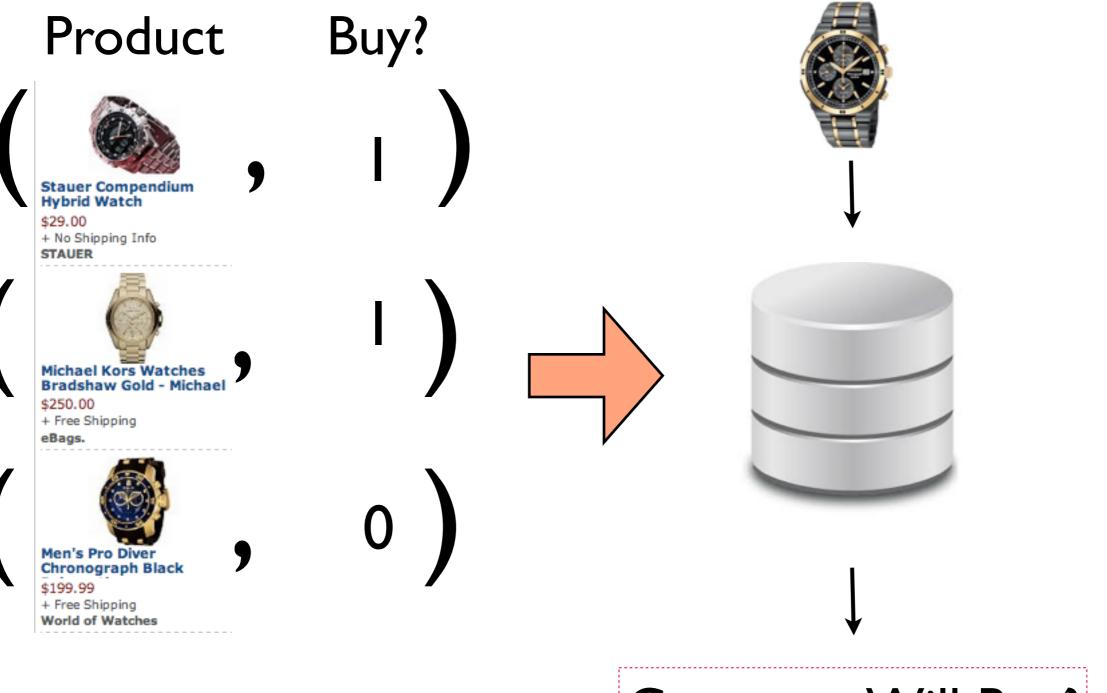




Sales Data



Sales Prediction Model



Customer Will Buy?

Zoolaster can potentially buy more watches from wholesaler that have higher potential of selling online

Data to Classification

Given a set of features

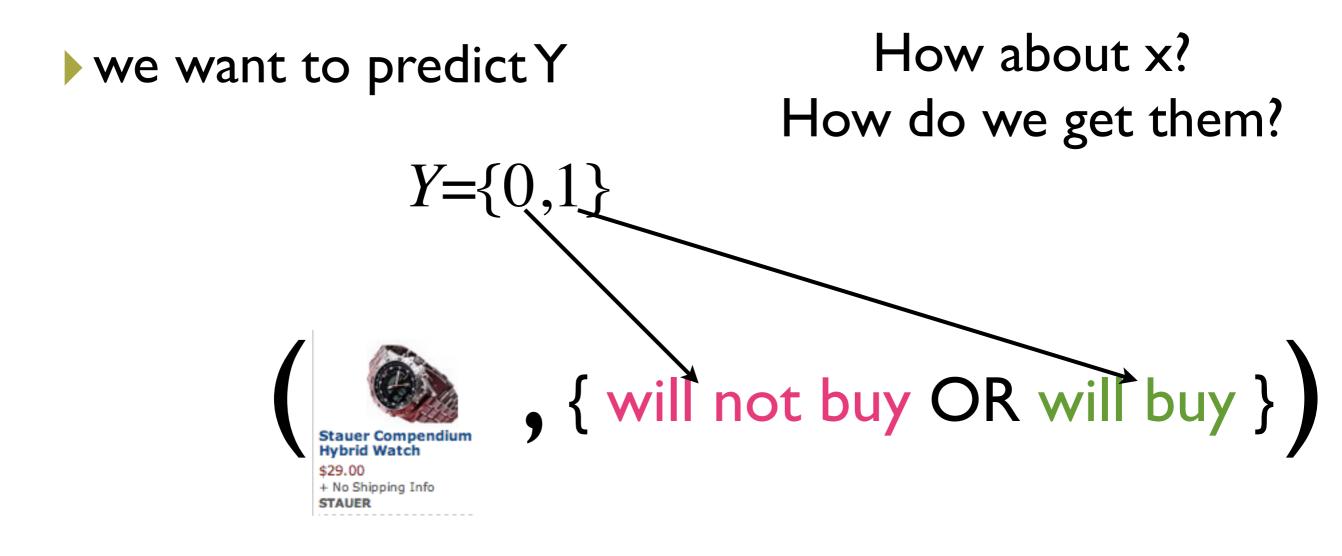
$$X = (x_1, x_2, x_3, \dots, x_n)$$

• we want to predict Y $Y=\{0,1\}$ How about x? How do we get them?

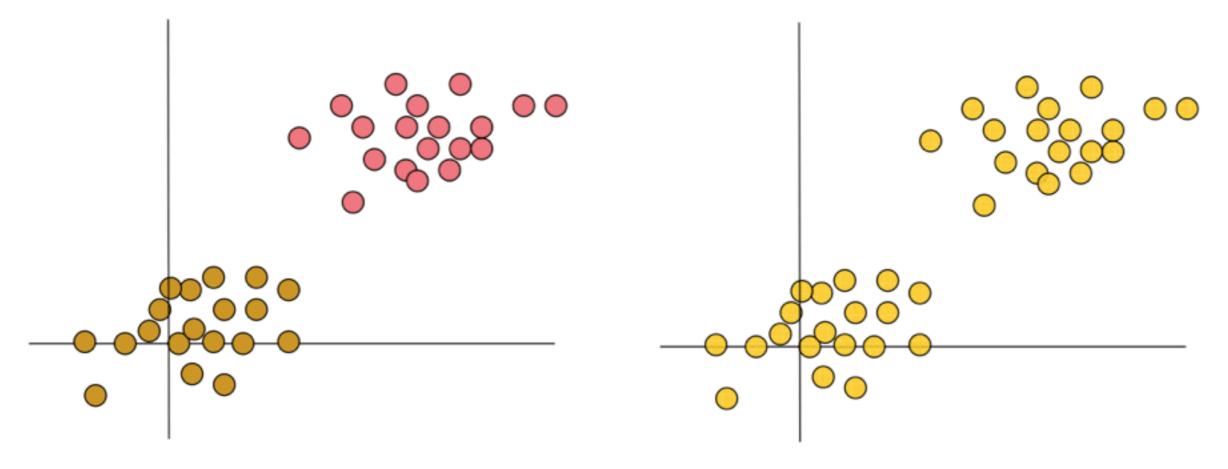
Data to Classification

Given a set of features

$$X = (x_1, x_2, x_3, \dots, x_n)$$



Data to Clusters



Supervised Training of Classification Algorithm Unsupervised Training of Clustering Algorithm