### CSE 158 — Lecture 1.5 Web Mining and Recommender Systems

#### Supervised learning – Regression

#### What is supervised learning?

### **Supervised learning** is the process of trying to infer from **labeled data** the underlying function that produced the labels associated with the data

#### What is supervised learning?

# Given labeled training data of the form $\{(\text{data}_1, \text{label}_1), \dots, (\text{data}_n, \text{label}_n)\}$

# Infer the function $f(\text{data}) \xrightarrow{?} \text{labels}$



### Suppose we want to build a movie recommender

#### e.g. which of these films will I rate highest?



#### Example

#### Q: What are the labels?

#### A: ratings that others have given to each movie, and that I have given to other movies

103 of 115 people found the following review helpful \*\*\*\*\*\* Excellent Sci-Fi, September 12, 2000 By Eric J. Pray (Upstate New York) - See all my reviews by <u>INF\_CATURE</u> (ubdate line interview): see and interview) The manual herm the line interview 
 Help other customers find the most helpful reviews
 Eccent abuse
 Pormains

 Was this review helpful to you? (Yes)
 No
 Comments (5)
 Comments (5)
 37 of 40 people found the following review helpful \*\*\*\*\* Sadly missed from the theatre. September 1, 2000 By Raja Chadni (Dehra Dun, India) - See al my reviews ew is from: Pitch Black (Unrated Version) (DVD) The control to the co Help other customers find the most helpful reviews
Was this review helpful to you? Yes No
Comment 18 of 19 people found the following review helpful Anthony Cool monster flick, November 14, 2000 By Kathy - See all my reviews Pitch Black [VHS] (VHS Tape) The new minimum register of the second secon 
 Help other customers find the most helpful reviews
 Report aburg
 Remaining

 Was this review helpful to you?
 Yes
 No.
 Comment
 83 of 101 people found the following review helpful www.initiation. Taut, smart, enjoyable filmmaking, November 14, 2000 By A Customer New is from: Pitch Black (Unrated Version) (DVD) 
 Help other customers find the most helpful reviews
 Report aburg
 Perms

 Was this review helpful to you? Yes
 No
 Comment (1)

18 of 20 people found the following review helpful schedule: A Slick Sci-Fi Thriller, May 3, 2001 By A. Philips - See all my reviews

#### Example

#### Q: What is the data?

## **A: features** about the movie and the users who evaluated it

#### Movie features: genre, actors, rating, length, etc.

#### **Product Details**

Genres	Science Fiction, Action, Horror
Director	David Twohy
Starring	Vin Diesel, Radha Mitchell
Supporting actors	Cole Hauser, Keith David, Lewis Fitz-Gerald, Claudia Black, Rhiana Gr Angela Moore, Peter Chiang, Ken Twohy
Studio	NBC Universal
MPAA rating	R (Restricted)
Captions and subtitles	English Details 🔻
Rental rights	24 hour viewing period. Details 🔻
Purchase rights	Stream instantly and download to 2 locations Details
Format	Amazon Instant Video (streaming online video and digital download)

User features: age, gender, A. Phillips location, etc.

Reviewer ranking: #17,230,554

90% helpful votes received on reviews (151 of 167)

ABOUT ME Enjoy the reviews...

#### ACTIVITIES

Reviews (16) Public Wish List (2) Listmania Lists (2) Tagged Items (1)



### Movie recommendation: $f(\text{data}) \xrightarrow{?} \text{labels}$

### $f(\text{user features}, \text{movie features}) \xrightarrow{?} \text{star rating}$

#### Solution 1

## Design a system based on **prior knowledge**, e.g.

```
def prediction(user, movie):
    if (user['age'] <= 14):
        if (movie['mpaa_rating']) == "G"):
            return 5.0
        else:
            return 1.0
    else if (user['age'] <= 18):
        if (movie['mpaa_rating']) == "PG"):
            return 5.0
..... Etc.
```

### Is this supervised learning?



Identify words that I frequently mention in my social media posts, and recommend movies whose plot synopses use **similar** types of language

Plot synopsis



to eat burritos for every meal iecember 22, 2014 at 10:42am - Like - 🖒 1

Vrite a comment

0

Sigh... I just had my muscles described as "not convincing" in the departmental newsletter. Time to go crawl into a hole and die I suppose.

#### **\*\*\*\*\*** 775 IMDb 7.1/10 **MITCH BLA** Is this supervised learning? in the shadows, waiting to attack in the dark, and the planet is rapidly plunging into the utter Computer Science and Er See More Starring: Vin Diesel, Radha Mitchell Runtime: 1 hour 53 minutes CSE Celebrates 2014 with Party, Festive Skits by Staff, Available to watch on supported devices. Students and Faculty | Computer Science... The 2014 end-of-vear department potluck holiday party (right) and CSE Holiday Skits took place Friday, December 12, and the mood was predictably festive. After the. CSE.UCSD.EDU Like - Comment - Share Michael Nguyen Taylor, Melanie Carmody, Javen Qinfeng Shi and 2 others like this. Katie Louise Down After you've eaten some more chicken breast. December 21, 2014 at 6:08pm · Unlike · 🖒 1 Melanie Carmody Oh no! what happened to the burrito diet? ecember 22, 2014 at 12:08am - Like Julian McAuley Unfortunately the trappings of adult life have made it impossible

argmax similarity(synopsis, post)

#### Solution 3

#### Identify which attributes (e.g. actors, genres) are associated with positive ratings. Recommend movies that exhibit those attributes.

### Is this supervised learning?

#### Solution 1

## (design a system based on prior knowledge)

Disadvantages:

- Depends on possibly false **assumptions** about how users relate to items
- Cannot adapt to new data/information Advantages:
- Requires no data!



## (identify similarity between wall posts and synopses)

Disadvantages:

- Depends on possibly false assumptions about how users relate to items
- May not be adaptable to new settings Advantages:
- Requires data, but does not require labeled data

#### Solution 3

## (identify attributes that are associated with positive ratings)

Disadvantages:

 Requires a (possibly large) dataset of movies with labeled ratings

Advantages:

- Directly optimizes a measure we care about (predicting ratings)
- Easy to adapt to new settings and data

#### Supervised versus unsupervised learning

### Learning approaches attempt to model data in order to solve a problem

**Unsupervised learning** approaches find patterns/relationships/structure in data, but **are not** optimized to solve a particular predictive task

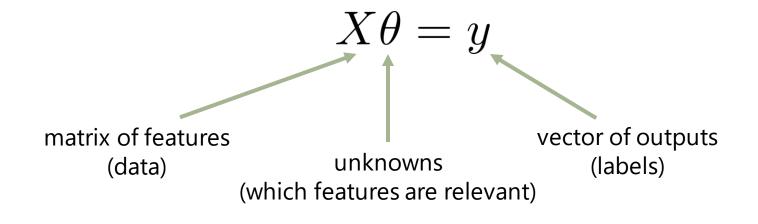
**Supervised learning** aims to directly model the relationship between input and output variables, so that the output variables can be predicted accurately given the input

#### Regression

**Regression** is one of the simplest supervised learning approaches to learn relationships between input variables (features) and output variables (predictions)

#### Linear regression

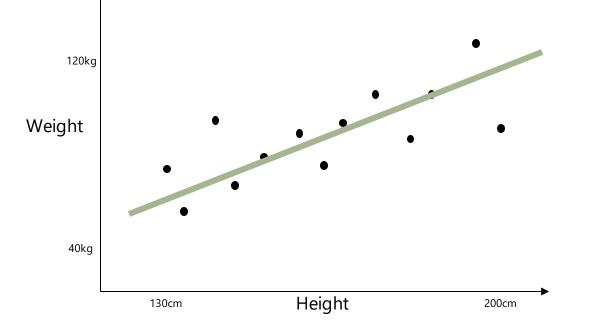
## **Linear regression** assumes a predictor of the form



(or Ax = b if you prefer)

#### Motivation: height vs. weight

**Q:** Can we find a line that (approximately) fits the data?



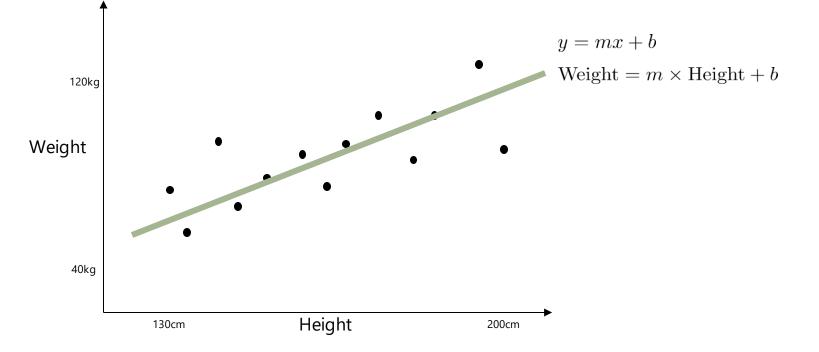
#### Motivation: height vs. weight

**Q:** Can we find a line that (approximately) fits the data?

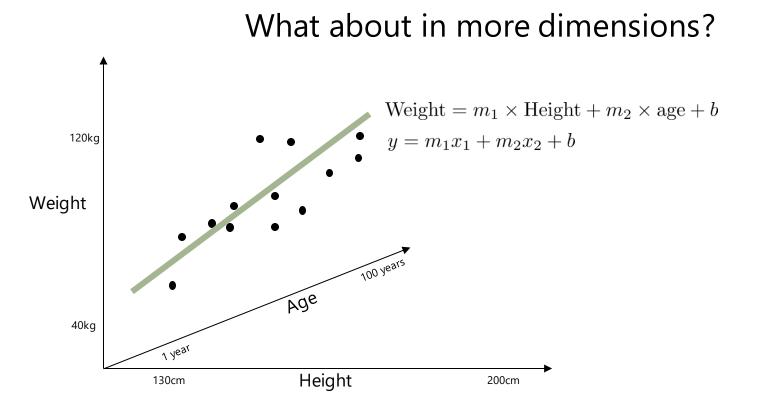
- If we can find such a line, we can use it to make **predictions** (i.e., estimate a person's weight given their height)
  - How do we **formulate** the problem of finding a line?
  - If no line will fit the data exactly, how to **approximate?** 
    - What is the "best" line?

#### Recap: equation for a line

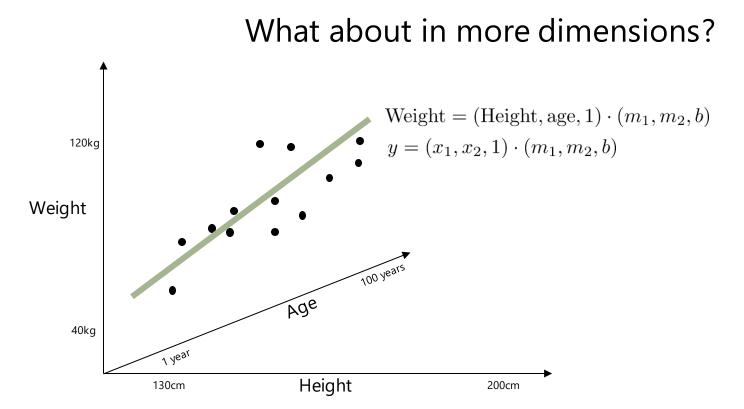
#### What is the formula describing the line?



#### Recap: equation for a line



### Recap: equation for a line as an inner product



#### Linear regression

## **Linear regression** assumes a predictor of the form

$$X\theta = y$$

**Q:** Solve for theta **A:** 



## How do preferences toward certain beers vary with age?

#### Example 1

### **Beeradvocate**

**Beers**:



world-class 9,587 Ratings (view ratings) Brewed by:

Goose Island Beer Co. Illinois, United States

BA SCORE

100

world-class

Style | ABV American Double / Imperial Stout | 13.80% ABV

THE BROS

95

Ratings: 9,587 Reviews: 2,537

rAvg: 4.59

pDev: 9.59% Wants: 2,109

Gots: 4,563 | FT: 472

Availability: Winter

Notes/Commercial Description: 60 IBU

(Beer added by: drewbage on 06-26-2003)

Displayed for educational use only; do not reuse.

#### **Ratings/reviews:**



4.35/5 rDev -5.2% look: 4 | smell: 4.25 | taste: 4.5 | feel: 4.25 | overall: 4.25

Serving: 355 mL bottle poured into a 9 oz Libbey Embassy snifter ("bottled on: 08AUG14 1109").

Appearance: Deep, dark near-black brown. Hazy, light brown fringe of foam and limited lacing; no head.

Smell: Roasted malt, vanilla, and some warming alcohol.

Taste: Roasted malts, cocoa, burnt caramel, molasses, vanilla and dark fruit. Bourbon barrel is hinted at but never takes over.

Mouthfeel: Medium to full body and light carbonation with a very lush, silky smooth feel.

Overall: Not as complex or intense as some newer barrel-aged stouts, but so smooth and balanced with all the elements tightly integrated.

#### HipCzech, Yesterday at 05:38 AM



	HipCzech Aficionado Male, from Texas Profile Page			$\otimes$
8	Member Since: Points: Beers: Places: Posts: Uikes Received: Trading:		HipCzech was last seen: Today at 12:19 AM	



50,000 reviews are available on http://jmcauley.ucsd.edu/cse158/data/beer/beer\_50000.json (see course webpage)

See also – non-alcoholic beers:

http://jmcauley.ucsd.edu/cse158/data/beer/non-alcoholic-beer.json



#### **Real-valued** features

### How do preferences toward certain beers vary with age? How about **ABV**?

(code for all examples is on <a href="http://jmcauley.ucsd.edu/cse158/code/week1.py">http://jmcauley.ucsd.edu/cse158/code/week1.py</a>)



#### **Real-valued** features

### What is the interpretation of: $\theta = (3.4, 10e^{-7})$

(code for all examples is on <a href="http://jmcauley.ucsd.edu/cse158/code/week1.py">http://jmcauley.ucsd.edu/cse158/code/week1.py</a>)

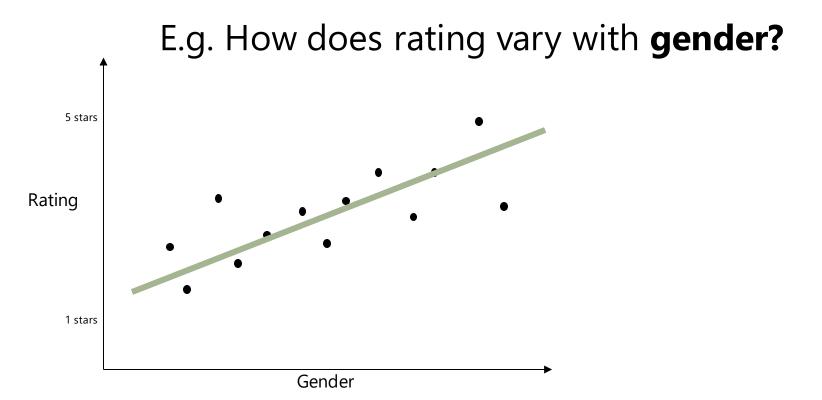


#### **Categorical features**

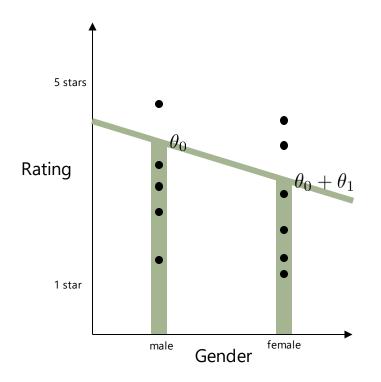
## How do beer preferences vary as a function of **gender**?

(code for all examples is on <a href="http://jmcauley.ucsd.edu/cse158/code/week1.py">http://jmcauley.ucsd.edu/cse158/code/week1.py</a>)

#### Example 2



#### Example 2



 $\theta_0$  is the (predicted/average) rating for males

 $\theta_1$  is the **how much higher** females rate than males (in this case a negative number)

We're really still fitting a line though!



#### Random features

### What happens as we add more and more **random** features?

(code for all examples is on http://jmcauley.ucsd.edu/cse158/code/week1.py)

#### Exercise

### How would you build a feature to represent the **month**, and the impact it has on people's rating behavior?