CSCE 689: Special Topics in Modern Algorithms for Data Science

Lecture 30

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Presentation Schedule

- November 27: Chunkai, Jung, Galaxy Al
- November 29: STMI, Anmol, Jason
- December 1: Bokun, Ayesha, Dawei, Lipai



d

n



- Let A be a set of n observations, each with d features
- Let *b* be the vector of outcomes/labels for each observation

• Find the vector x such that Ax = b



Karl Rohe @karlrohe · 3h i'm teaching linear regression in the spring

First thing we'll do: highlight the two *very* different motivations for linear models

for causal inference
for prediction

Breiman's "two cultures" paper is too advanced for midundergrads... are there any easier readings?

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- Economics and Finance:
 - Stock price prediction: Predict stock prices based on historical data and other relevant factors.
 - Econometric analysis: Study the relationships between economic variables like GDP, inflation, and interest rates.





- Medicine and Healthcare:
 - Medical research: Understand the relationship between factors such as age, genetics, and lifestyle on health outcomes.
 - Disease prediction: Predicting the probability of disease occurrence based on risk factors like smoking, diet, and exercise.



- Sports Analytics:
 - Player performance forecast: Predicting player performance in sports like baseball, basketball, or soccer based on historical data.
 - Game outcome prediction: Predicting the outcome of games based on team statistics and other factors.





- Environmental Science:
 - Climate modeling: Model climate change variables like temperature, precipitation, and sea levels.
 - Pollution analysis: Analyze the relationship between pollution levels and various factors like industrial activity and population.





- Let A be a set of n observations, each with d features
- Let *b* be the vector of outcomes/labels for each observation

• Find the vector x such that Ax = b



• What are conditions for finding *x* such that Ax = b?



- What are conditions for finding *x* such that Ax = b?
- If the system is inconsistent, there are no solutions



- What are conditions for finding x such that Ax = b?
- If the system is inconsistent, there are no solutions
- Can check for consistency by looking at the rank of the coefficient matrix and the augmented matrix

• For a square matrix $A \in \mathbb{R}^{n \times n}$, if A is full rank, i.e., rank(A) = n, then A has an inverse $A^{-1} \in \mathbb{R}^{n \times n}$ such that $AA^{-1} = A^{-1}A = \mathbb{I}_n$

• For a general matrix $A \in \mathbb{R}^{n \times d}$ with linearly independent columns, the Moore-Penrose inverse/pseudoinverse A^{\dagger} of A satisfies $A^{\dagger} = (A^{\top}A)^{-1}A^{\top}$ so that $A^{\dagger}A = \mathbb{I}_d$



- What are conditions for finding x such that Ax = b?
- If the system is inconsistent, there are no solutions
- If the system is consistent and n = d, then there is a single solution, $x = A^{-1}b$



- What are conditions for finding x such that Ax = b?
- If the system is inconsistent, there are no solutions
- If the system is consistent and *n* < *d*, then are infinite solutions and can find a solution by looking at *n* columns of *A*



- What are conditions for finding x such that Ax = b?
- If the system is inconsistent, there are no solutions
- If the system is consistent and *n* > *d*, then at most one solution (*x* = *A*[†]*b*) and can find by looking at *d* rows of *A*



- What are conditions for finding x such that Ax = b?
- What to do when there is no x such that Ax = b?



- What are conditions for finding x such that Ax = b?
- What to do when there is no x such that Ax = b?
- Minimize $\mathcal{L}(Ax b)$ for some loss function \mathcal{L}



- Find the vector x that minimizes $||Ax b||_2$
- "Least squares" optimization





- Find the vector x that minimizes $||Ax b||_2$
- What is the solution to

 $\min_{x\in\mathbb{R}^d}\|Ax-b\|_2?$



Find x that minimizes $||Ax - b||_2$

 $b = \begin{bmatrix} 5 \\ 5 \end{bmatrix}$









- Find the vector x that minimizes $||Ax b||_2$
- What is the solution to $\min_{x \in \mathbb{R}^d} \|Ax - b\|_2?$
- Let b^{\perp} be the projection of bonto the column space of A
- Decompose b into b^{\perp} and b^{\parallel}

• We have $\arg\min_{x\in\mathbb{R}^d} ||Ax - b||_2 = \arg\min_{x\in\mathbb{R}^d} ||Ax - b||_2^2$

•
$$||Ax - b||_2^2 = ||Ax - b^{\perp} - b^{\parallel}||_2^2$$

• We have $\arg\min_{x\in\mathbb{R}^d} ||Ax - b||_2 = \arg\min_{x\in\mathbb{R}^d} ||Ax - b||_2^2$

•
$$||Ax - b||_2^2 = ||Ax - b^{\perp} - b^{\parallel}||_2^2$$

= $||Ax - b^{\parallel}||_2^2 - 2\langle Ax - b^{\parallel}, b^{\perp} \rangle + ||b^{\perp}||_2^2$

• We have $\arg\min_{x\in\mathbb{R}^d} ||Ax - b||_2 = \arg\min_{x\in\mathbb{R}^d} ||Ax - b||_2^2$

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= $||Ax - b^{\parallel}||_2^2 - 2\langle Ax - b^{\parallel}, b^{\perp} \rangle + ||b^{\perp}||_2^2$
= $||Ax - b^{\parallel}||_2^2 + ||b^{\perp}||_2^2$

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$$||Ax - b||_2^2 = ||Ax - b^{\perp} - b^{\parallel}||_2^2$$

= $||Ax - b^{\parallel}||_2^2 - 2\langle Ax - b^{\parallel}, b^{\perp} \rangle + ||b^{\perp}||_2^2$
= $||Ax - b^{\parallel}||_2^2 + ||b^{\perp}||_2^2$

• Minimized for $||Ax - b^{\parallel}||_2^2 = 0$ when $x = A^{\dagger}b^{\parallel} = A^{\dagger}b$



- Find the vector x that minimizes $||Ax b||_2$
- "Least squares" optimization
- MLE under Gaussian noise
- Closed form solution: $x = A^{\dagger}b$

