

# Data Mining 2026 Spring!

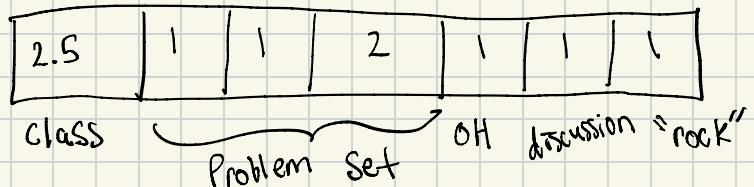
This class:

- linear algebra + probability
- machine learning
- foundations of gen AI

challenging course!

Prereqs: linear algebra, probability, algorithms, calc 3

Time Expectation: 12.5 hr/week



- Earlier = better
- More drawing = better
- More talking = better

[www.rtealwitter.com/datamining2026spring](http://www.rtealwitter.com/datamining2026spring)

- discord for communication
- reading for each lecture
- these slides are online

Plans

- Warm up
  - ↳ math review
  - ↳ Page Rank
- Supervised Learning
- Unsupervised Learning

## Math Review

machine learning?

calculus and linear algebra!

### Derivatives

$$f: \mathbb{R} \rightarrow \mathbb{R}$$

"maps real number to real number"

$$f'(x) = \frac{\partial}{\partial x} f(x)$$

$$= \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}$$

Function:  $f(x)$

$$x^2$$

$$x^a$$

$$ax + b$$

$$\ln(x)$$

$$e^x$$

Q: If my investment earns 10% per year, how much more will I have in 60 years?

Derivative:  $\frac{\partial}{\partial x} f(x)$

$$2x$$

In ML, we'll combine many functions (e.g., neural net)

$$f: \mathbb{R} \rightarrow \mathbb{R} \quad g: \mathbb{R} \rightarrow \mathbb{R}$$

$$f(x) = \ln(x) \quad g(x) = x^2$$

### Chain Rule

$$\frac{\partial [g(f(x))]}{\partial x} = g'(f(x)) f'(x)$$

$$=$$

### Product Rule

$$\frac{\partial [g(x) \cdot f(x)]}{\partial x} = g(x) f'(x) + f(x) g'(x)$$

$$=$$

### Gradients

In ML, process high-dim input

$$f: \mathbb{R}^d \rightarrow \mathbb{R}$$

= maps  $d$  real numbers to one real "

$f(x_1, x_2, \dots, x_d)$   $\leftarrow$  visualize as mountain with

$$= f\left(\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_d \end{bmatrix}\right) = f(\underline{x})$$

$\frac{\partial f(\underline{x})}{\partial x_i}$  = how changing  $x_i$  changes  $f$

$$= \lim_{h \rightarrow 0} \frac{f(x_1, \dots, x_i + h, \dots, x_d) - f(x_1, \dots, x_i, \dots, x_d)}{h}$$

$$\nabla_{\underline{x}} f(\underline{x}) = \begin{bmatrix} \frac{\partial f(\underline{x})}{\partial x_1} \\ \vdots \\ \frac{\partial f(\underline{x})}{\partial x_d} \end{bmatrix}$$

## Vector and Matrix Multiplication

\* crucial to ML, neural nets only succeeded with GPUs

$$\underline{u} \in \mathbb{R}^d$$

$$\begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_d \end{bmatrix}$$

$$\underline{v} \in \mathbb{R}^d$$

$$\begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_d \end{bmatrix}$$

$$\underline{u} \cdot \underline{v} = \sum_{i=1}^d u_i v_i = u_1 v_1 + u_2 v_2 + \dots + u_d v_d$$

$$= \underline{u}^T \underline{v} = [u_1 \ u_2 \ \dots \ u_d] \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_d \end{bmatrix}$$

$$= \underline{v}^T \underline{u} = [v_1 \ v_2 \ \dots \ v_d] \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_d \end{bmatrix}$$

$$= \langle \underline{u}, \underline{v} \rangle$$

## $\ell_2$ -norm

$$\|\underline{u}\|_2^2 = \underline{u} \cdot \underline{u} = \sum_{i=1}^d u_i^2$$

$$A \in \mathbb{R}^{d \times m}$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1m} \\ a_{21} \\ \vdots \\ a_{d1} \end{bmatrix}_{d \times m}$$

$$B \in \mathbb{R}^{m \times n}$$

$$\begin{bmatrix} b_{11} & \dots & b_{1n} \\ b_{21} \\ \vdots \\ b_{m1} \end{bmatrix}_{m \times n}$$

$$AB$$

$\underbrace{dxm \ m \times n}_{dxn}$

$$[AB]_{ij} = \sum_{k=1}^m a_{ik} b_{kj}$$

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \quad B = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$$

$$AB = ?$$

$$BA = ?$$