

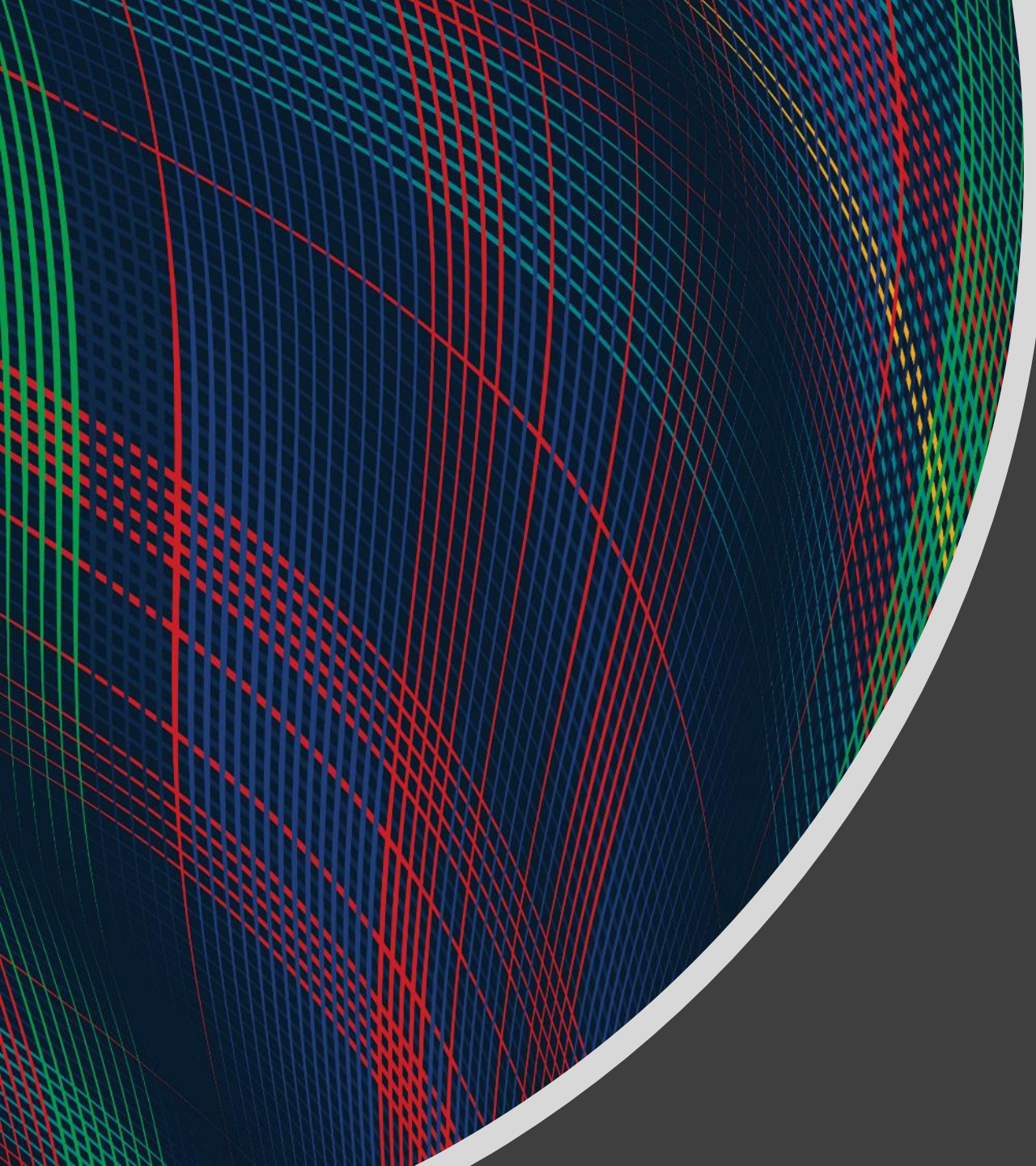
# Announcements

## HW2

- Due Sat 10/2

## Quizzes

- Mon 10/4, last 15 min. of class (calculus, optimization, Lagrange)
- Mon 10/11, last 15 min. of class (probability, statistics)



# Mathematical Foundations for Machine Learning

## Probability

Instructor: Pat Virtue

# Plan

## Today

### Probability

- Probability exercises
- Chain rule
- Independence
- i.i.d.

### Statistics

- Finding the best parameters

Probability

# Poll 1: Exercise

Implement a function in Python for the pdf of a Gaussian distribution.

Python numpy or math packages are fine, no scipy, etc.

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{\frac{-(x-\mu)^2}{2\sigma^2}}$$

# Probability Vocab

Outcomes

Sample space

Events

Probability

Random variable

Discrete random variable

Continuous random variable

Probability mass function

Probability density function

Parameters

# Probability Toolbox

- Algebra
- Three axioms of probability
- Theorem of total probability
- Definition of conditional probability
- Product rule
- Bayes' theorem
- Chain rule
- Independence
- Conditional independence

# Tools Summary

Adding to our toolbox

1. Definition of conditional probability
2. Product Rule
3. Bayes' theorem
4. Chain Rule...

$$P(A|B) = \frac{P(A, B)}{P(B)}$$

$$P(A, B) = P(A|B)P(B)$$

$$P(B|A) = \frac{P(A|B)P(B)}{P(A)}$$



# The Chain Rule

More generally, can always write any joint distribution as an incremental product of conditional distributions

# Independence

# Conditional Independence

# Probability Notation

Notation doc

Independent and identically distributed

# Likelihood

**Likelihood:** The probability (or density) of random variable  $Y$  taking on value  $y$  given the distribution parameters,  $\theta$ .

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Grades

Gaussian PDF: 
$$p(y \mid \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(y-\mu)^2}{2\sigma^2}}$$

# Likelihood

Trick coin: comes up heads only  $\frac{1}{3}$  of the time

1 flip: H	probability: $\frac{1}{3}$
2 flips: H,H	probability: $\frac{1}{3} \cdot \frac{1}{3}$
3 flips: H,H,T	probability: $\frac{1}{3} \cdot \frac{1}{3} \cdot \left(1 - \frac{1}{3}\right)$

But why can we just multiply these?



# Likelihood and i.i.d

**Likelihood:** The probability (or density) of random variable  $Y$  taking on value  $y$  given the distribution parameters,  $\theta$ .

**i.i.d.:** Independent and identically distributed

# Bernoulli Likelihood

Bernoulli distribution:

$$Y \sim \text{Bern}(\phi) \quad p(y \mid \phi) = \begin{cases} \phi, & y = 1 \\ 1 - \phi, & y = 0 \end{cases}$$

What is the likelihood for three i.i.d. samples, given parameter  $\phi$ :

$$\mathcal{D} = \{y^{(1)} = 1, y^{(2)} = 1, y^{(3)} = 0\}$$

$$\begin{aligned} & \prod_{i=1}^N p(Y = y^{(i)} \mid \phi) \\ &= \phi \cdot \phi \cdot (1 - \phi) \end{aligned}$$

## Poll 2

Assume that exam scores are drawn independently from the same Gaussian (Normal) distribution.

Given three exam scores 75, 80, 90, which pair of parameters is a better fit (a higher likelihood)?

- A) Mean 80, standard deviation 3
- B) Mean 85, standard deviation 7

Use a calculator/computer.

Gaussian PDF: 
$$p(y \mid \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(y-\mu)^2}{2\sigma^2}}$$