

# Probability Distributions

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# Introduction



# Session Objective

1. Define the fundamental continuous and discrete probability distributions
2. Use matplotlib to visualize distributions of data and discuss data vis fundamentals



# Resource



- “Think Stats” is available free online, or through amazon (I purchased a copy, I like authors)
- Great resource if any of the material today goes by too fast

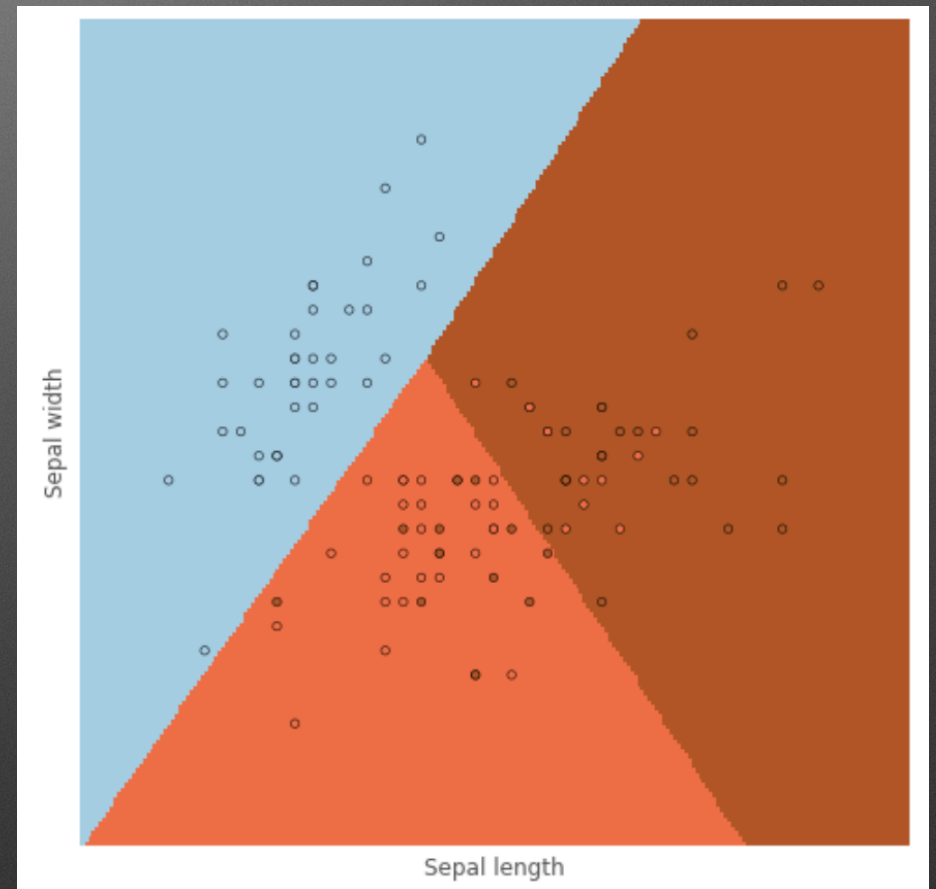


# Basics of Data Visualization



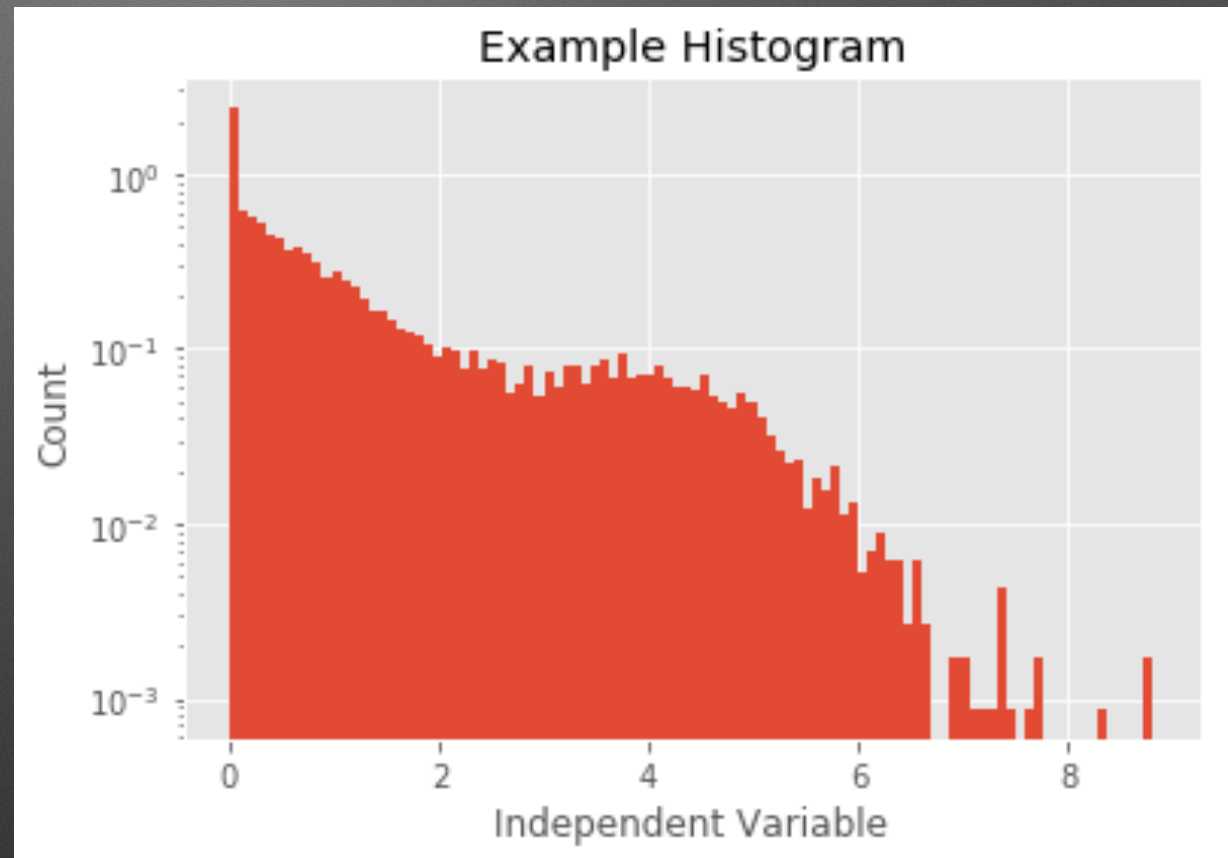
# Data Visualization

- Data visualization is a skill that you only get better at through thoughtful practice
- Depicted is the predictions of a logistic regression model on the iris dataset



# Histograms

- The histogram is one of the more fundamental depictions of data.
- Histograms depict the how often a certain value occurs in a set of data.
- Let's hop into a notebook and go over some basic histogram creation.





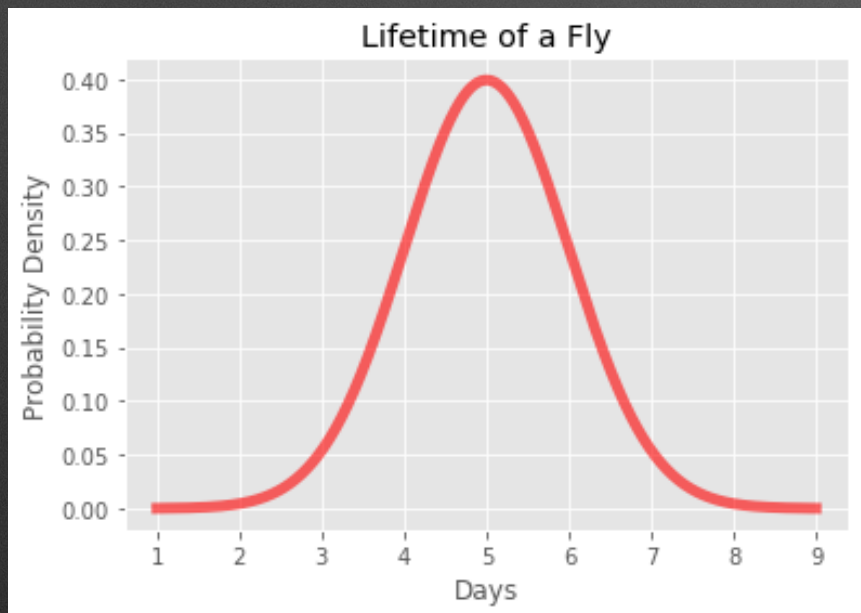
# Fundamentals of Distributions



# Continuous vs. Discrete

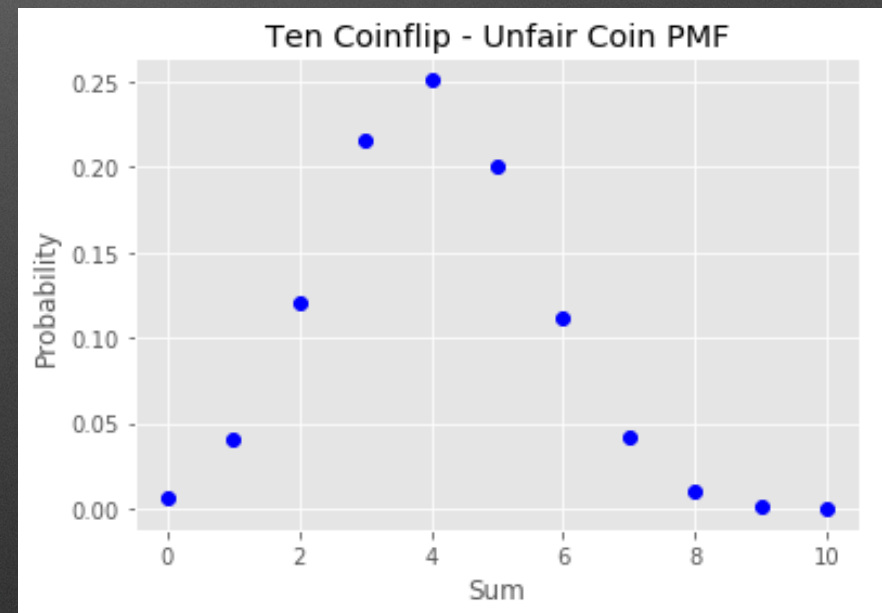
## Continuous

- floating point numbers
- the life span of a fly
- prob. at a point is 0, only an integral has non-zero probability



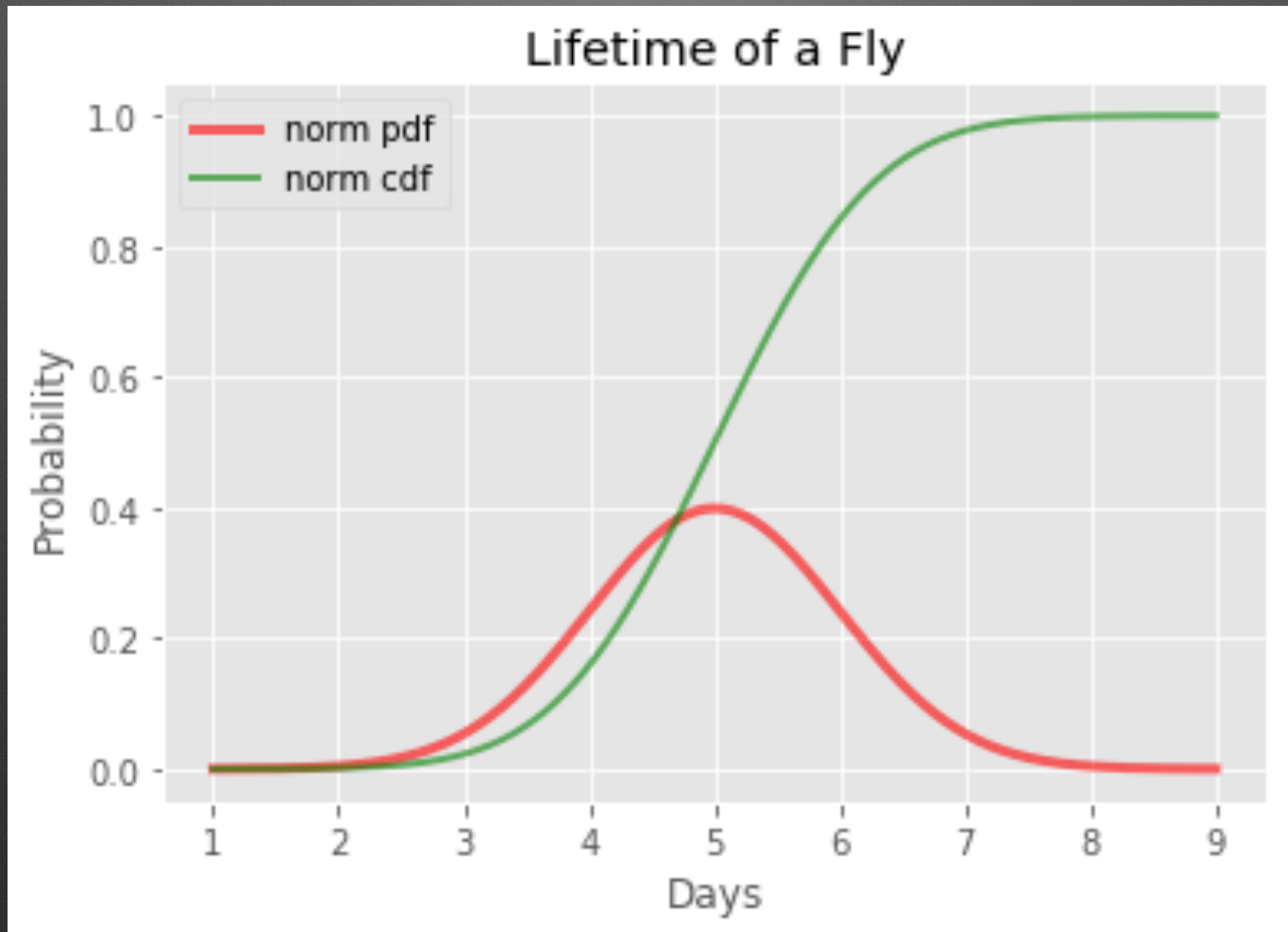
## Discrete

- integers
- e.g. the number of heads when flipping coins



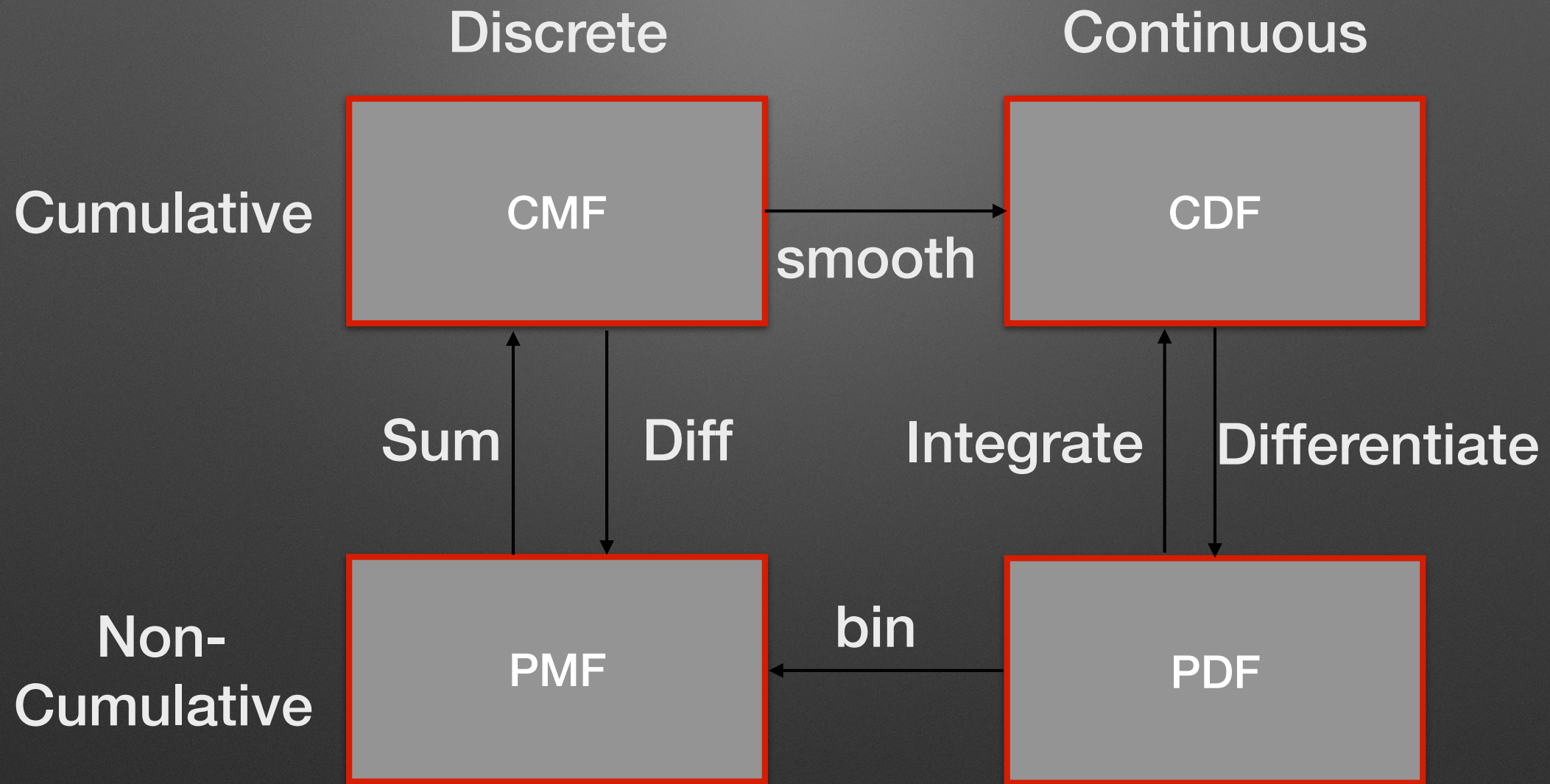


# Cumulative Distribution Functions





# Relationships





# Expectation & Variance

## Recall: Expectation and Variance

For **discrete** random variables (let  $P$  be the PMF of the r.v.  $X$ ):

$$E(X) = \sum_{s \in S} s * P(X = s)$$

$$Var(X) = \sum_{s \in S} (s - E(X))^2 * P(X = s)$$

For **continuous** random variables (let  $f$  is the PDF of r.v.  $X$ ):

$$E(X) = \int_{x=-\infty}^{\infty} x * f(x) dx$$

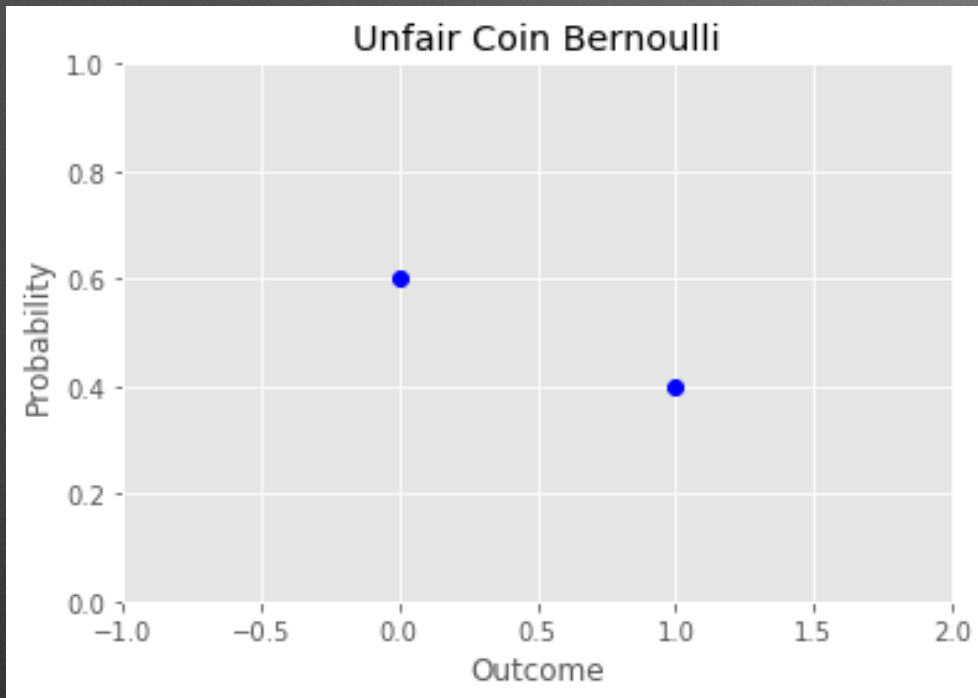
$$Var(X) = \int_{x=-\infty}^{\infty} (x - E(X))^2 * f(x) dx$$



# Important Discrete Distributions



# Bernoulli & Binomial

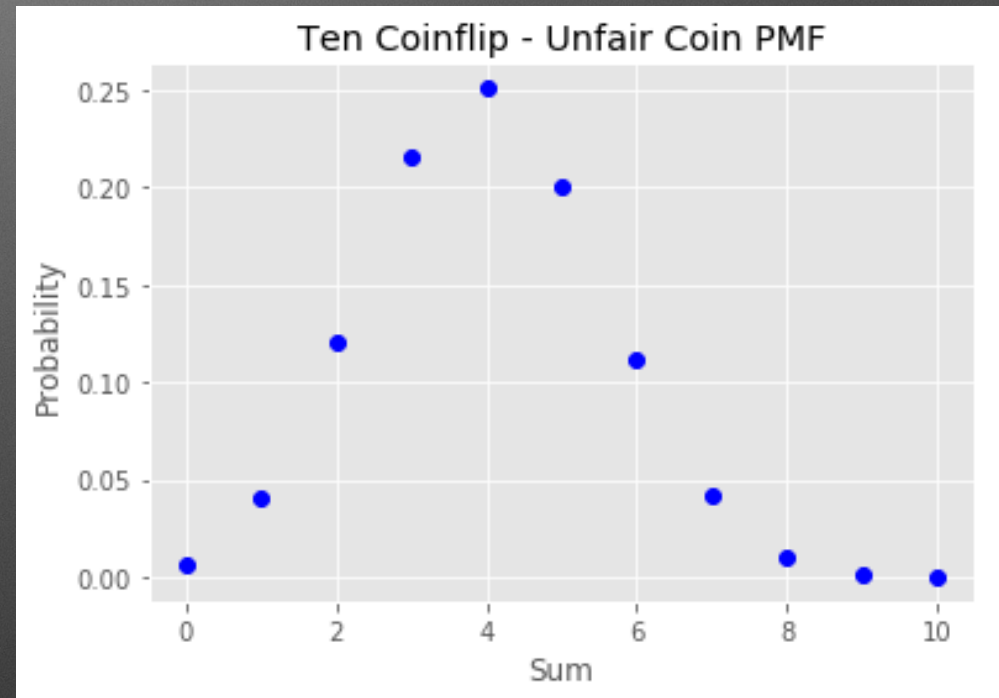


PMF:  $P[\text{success}] = p$ ,  $P[\text{failure}] = 1 - p$

Support:  $\{\text{success}, \text{failure}\}$

Mean:  $p$

Variance:  $p(1 - p)$



PMF:  $P[X = k] = \binom{n}{k} p^k (1 - p)^{n-k}$

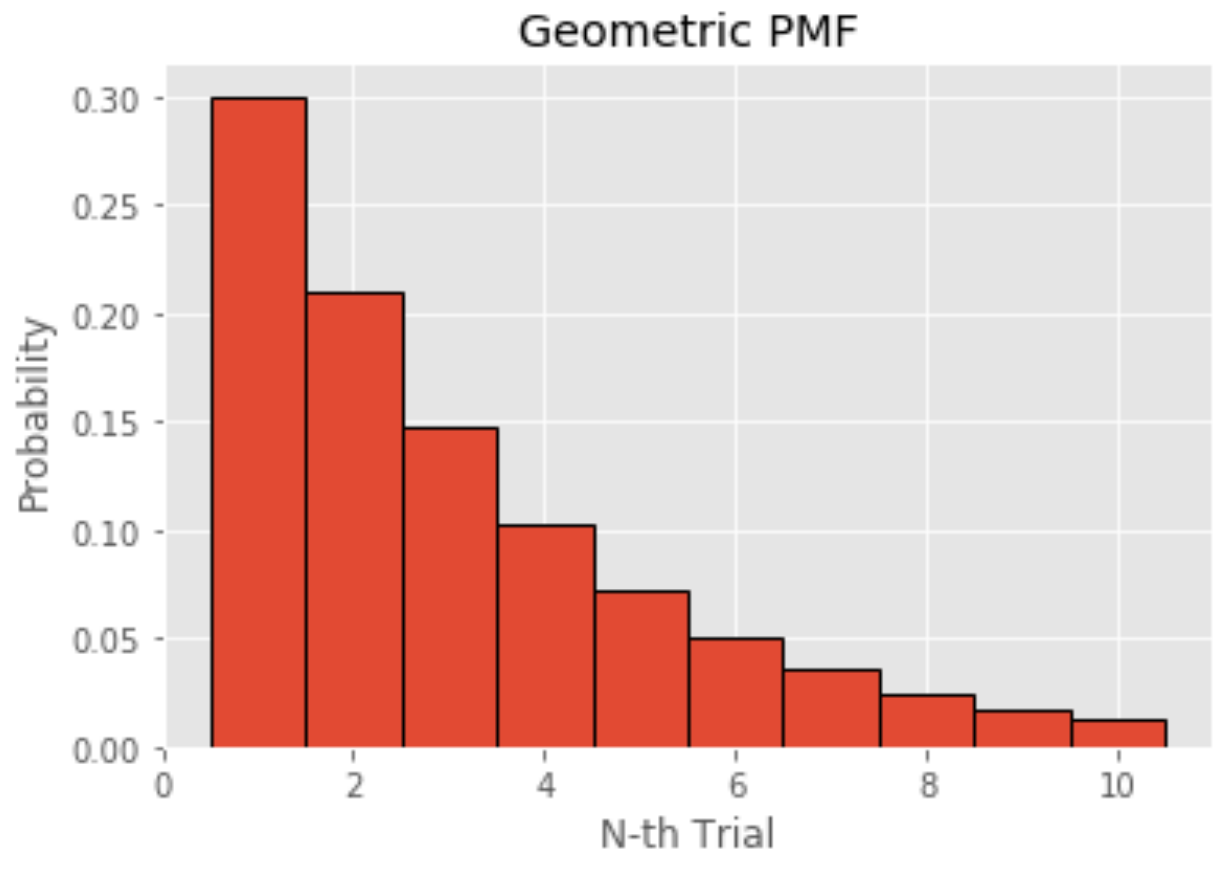
Support:  $k \in \{0, 1, \dots, n\}$

Mean:  $np$

Variance:  $np(1 - p)$



# Geometric Distribution



PMF:  $P[X = k] = p(1 - p)^{k-1}$

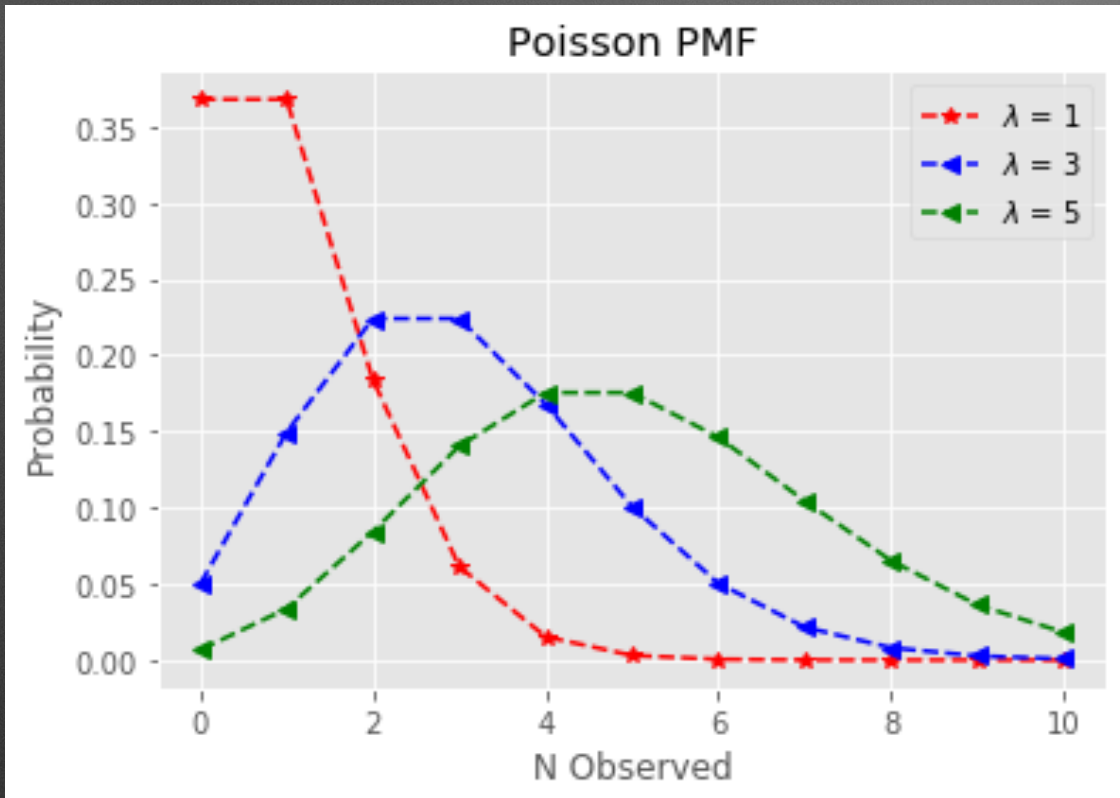
Support:  $k \in \{0, 1, \dots\}$

Mean:  $\frac{1}{p}$

Variance:  $\frac{1-p}{p^2}$



# Poisson Distribution



$$\text{PMF: } P[X = k] = \frac{\lambda^k e^{-\lambda}}{k!}$$

Support:  $k \in \{0, 1, 2, \dots\}$

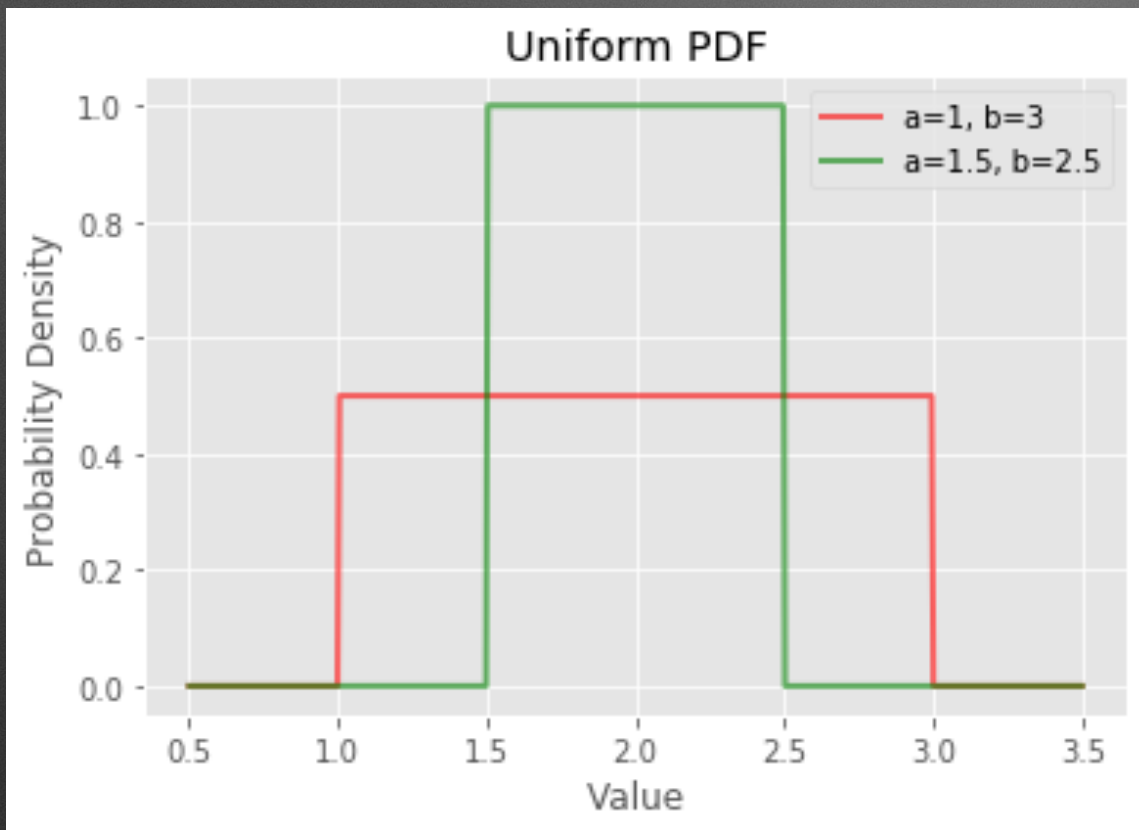
Mean:  $\lambda$

Variance:  $\lambda$

# Important Continuous Distributions



# Uniform Distribution



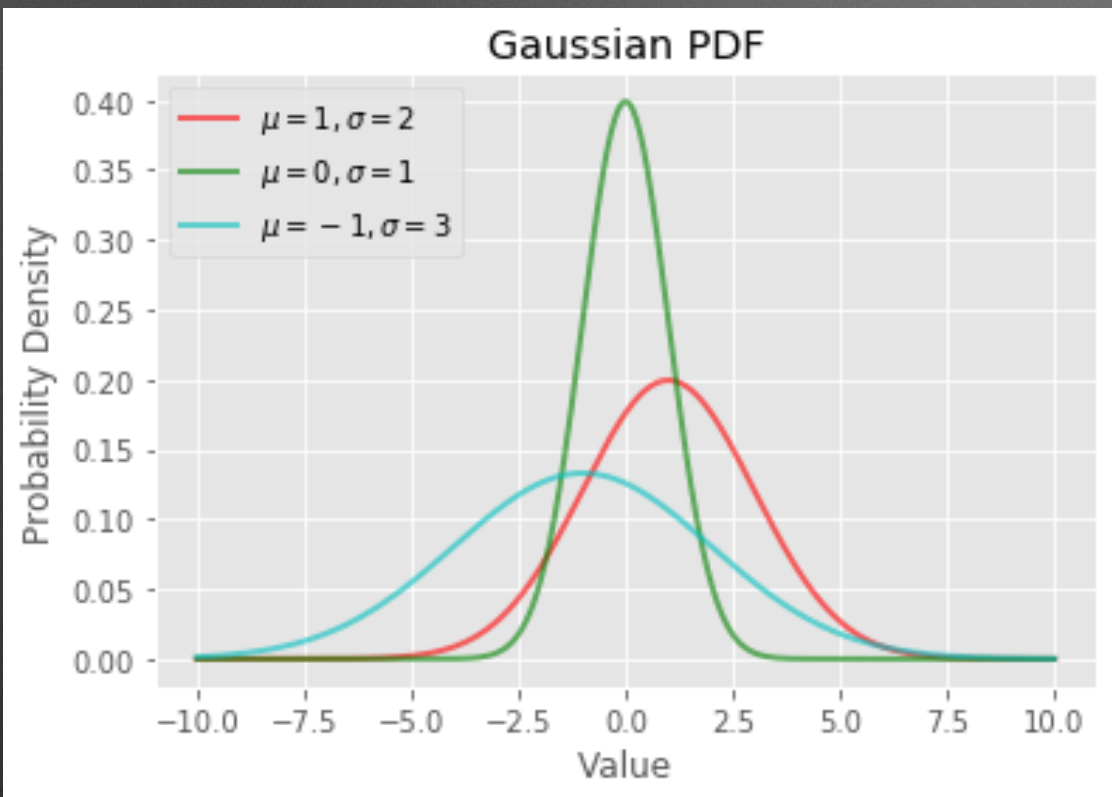
$$\text{PDF: } f(x) = \frac{1}{b-a} * [\theta(a-x) - \theta(b+a-x)]$$

$$\text{Support: } x \in [a, b]$$

$$\text{Mean: } \frac{a+b}{2}$$

$$\text{Variance: } \frac{(b-a)^2}{12}$$

# Normal Distribution



$$\text{PDF: } f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$$

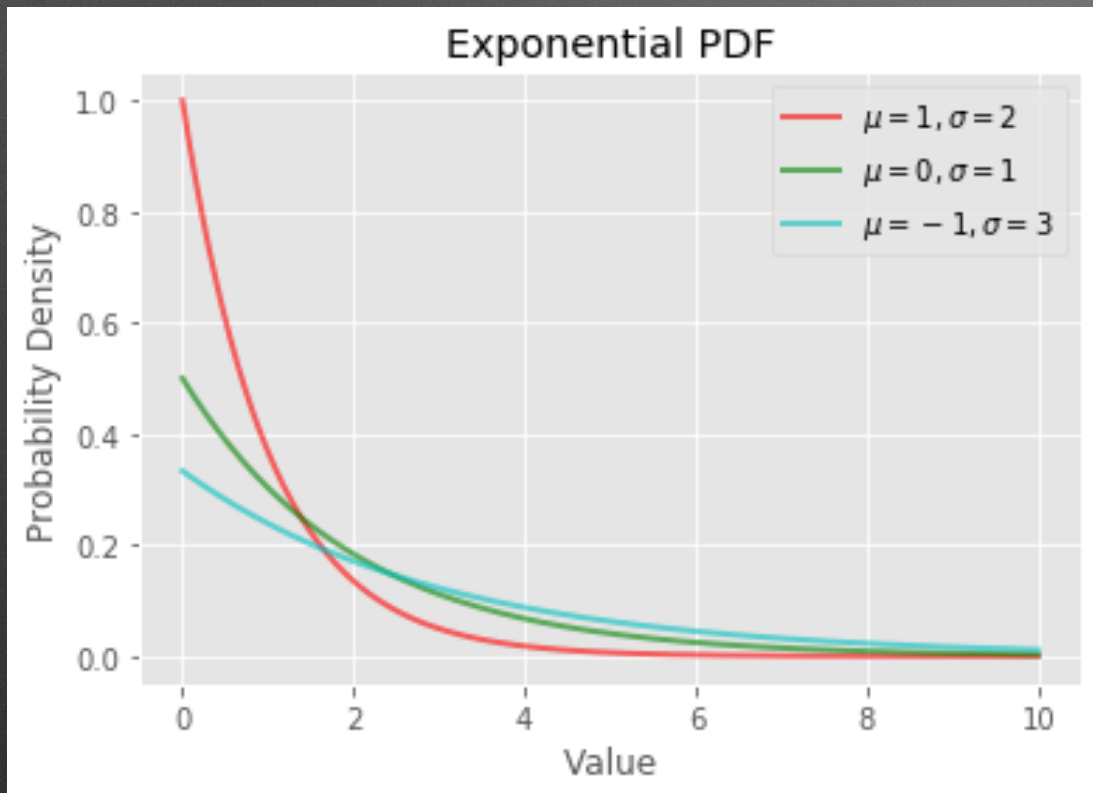
Support:  $x \in (-\infty, \infty)$

Mean:  $\mu$

Variance:  $\sigma^2$



# Exponential Distribution



PDF:  $f(x) = \lambda \exp(-\lambda x)$

Support:  $x \in [0, \infty)$

Mean:  $\frac{1}{\lambda}$

Variance:  $\frac{1}{\lambda^2}$