

# // section 12: statistical distributions

## stuff to learn today:

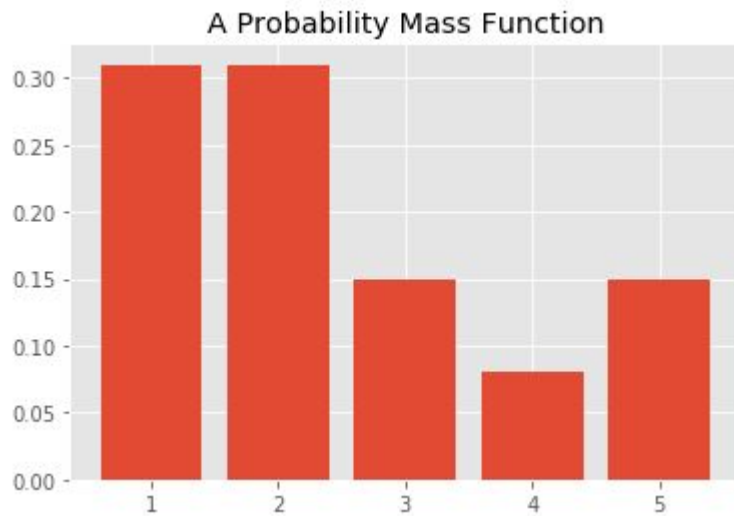
1.  $E(X)$  &  $\text{Var}(X)$
2. PMF, PDF, CDF
3. Bernoulli & Binomial Distributions
4. Normal Distributions
5. One Sample Z-Test

## expected value & variance

- for a **random variable**:
  - expected value =  $E(X) = \sum p(x_i) \cdot x_i$
  - variance =  $\text{Var}(X) = \sum p(x_i) \cdot (x_i - E(X))^2$
- for specific named distributions, there are often formulas for the expected value and variance

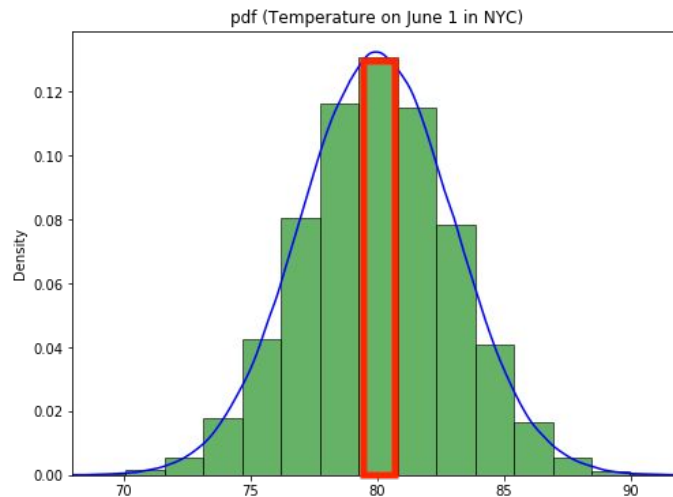
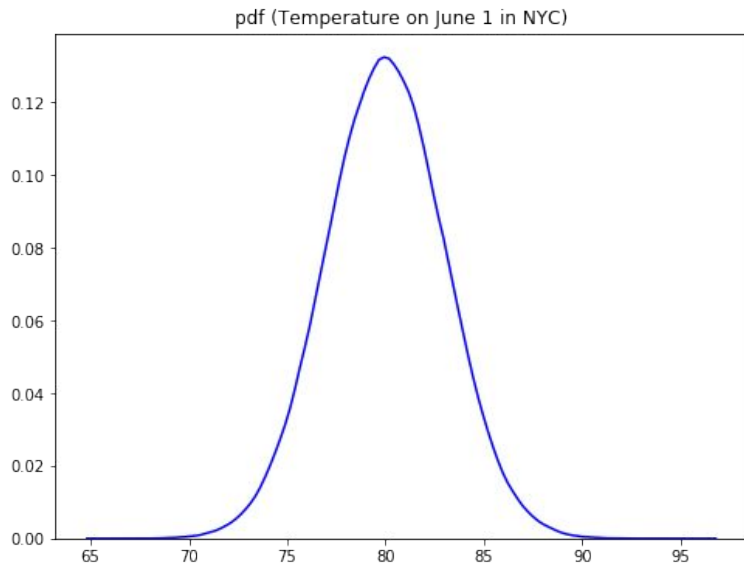
# probability mass function

- associates probabilities with discrete random variables
- discrete = a known number of possible outcomes



# probability mass function

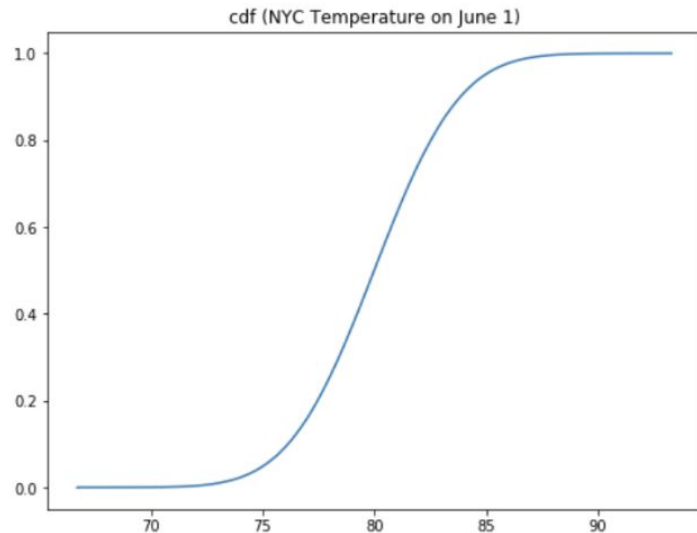
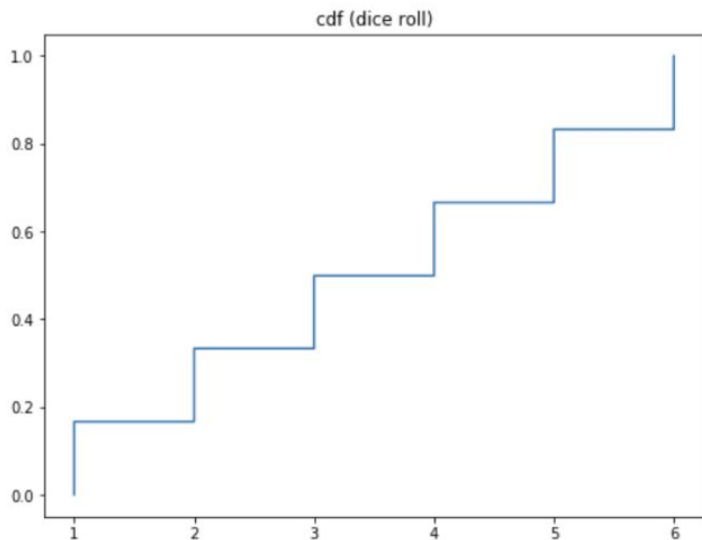
- associates probabilities with **continuous** random variables



13% of the time you'll observe a temperature between 79.3 and 80.8 degrees

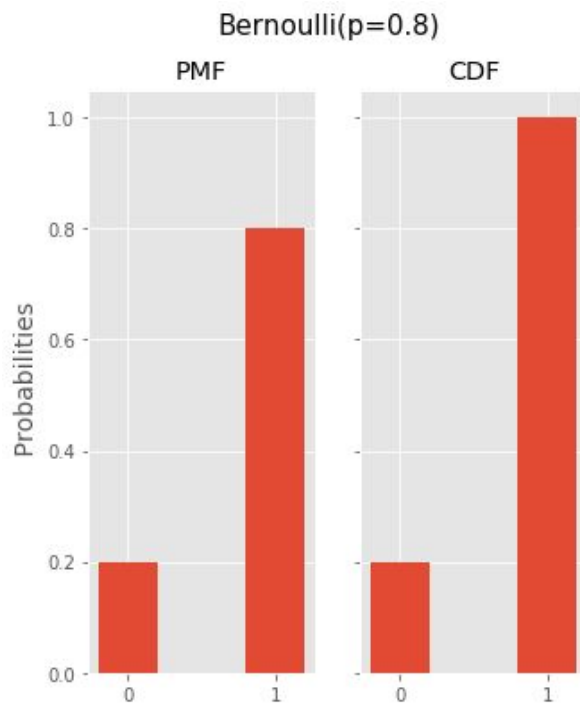
# cumulative distribution function

- shows  $P(X \leq x)$  for any  $x$  within the sample space



# the bernoulli distribution

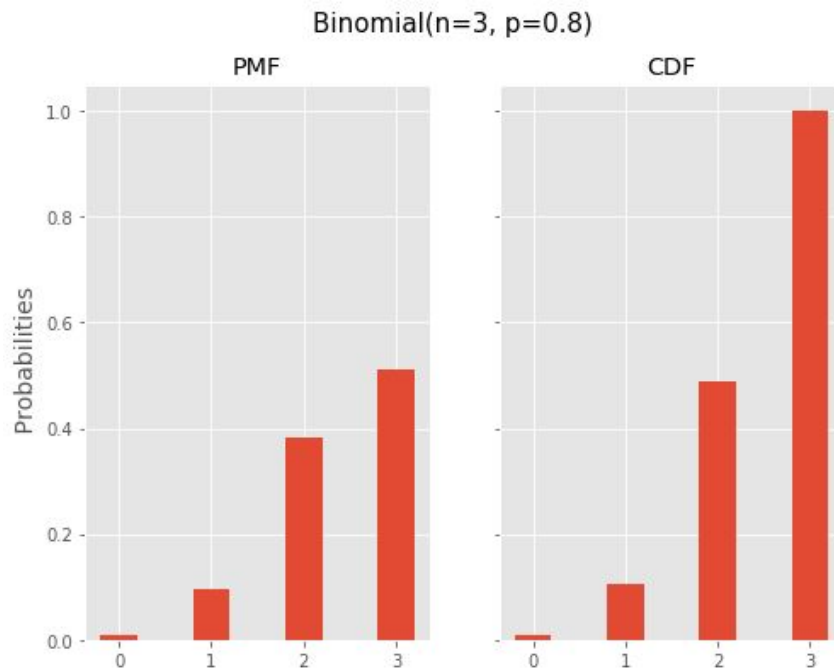
- a one-trial, binary outcome experiment
- $X \sim \text{Ber}(p=0.8)$
- $E(X) = 0.8$
- $\text{Var}(X) = 0.16$



# the binomial distribution

- a multi-trial, binary outcome experiment

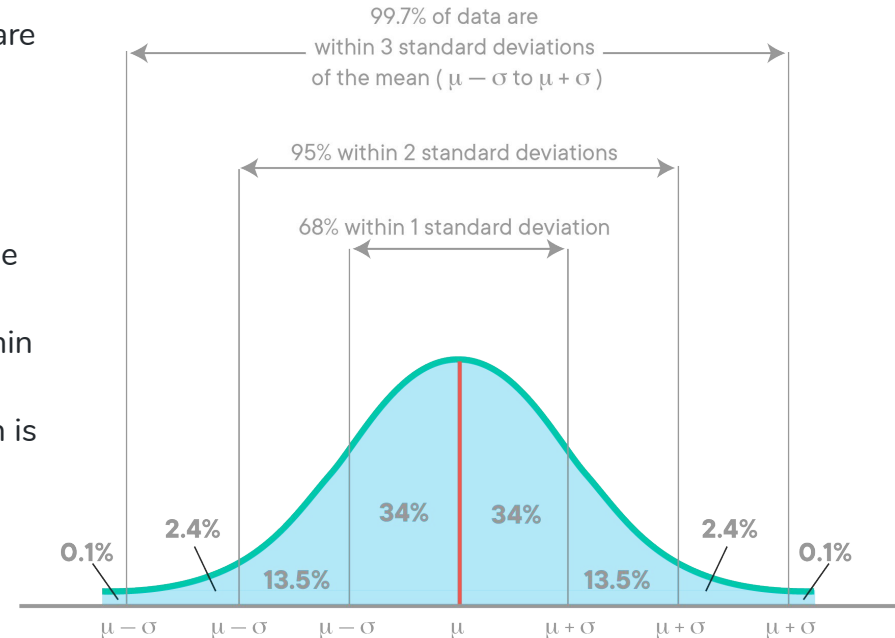
- $X \sim \text{Bi}(n = 3, p=0.8)$
- $P(X=k) = {}^nC_k (p)^k (1-p)^{n-k}$
- $E(X) = 2.4$
- $\text{Var}(X) = 0.48$



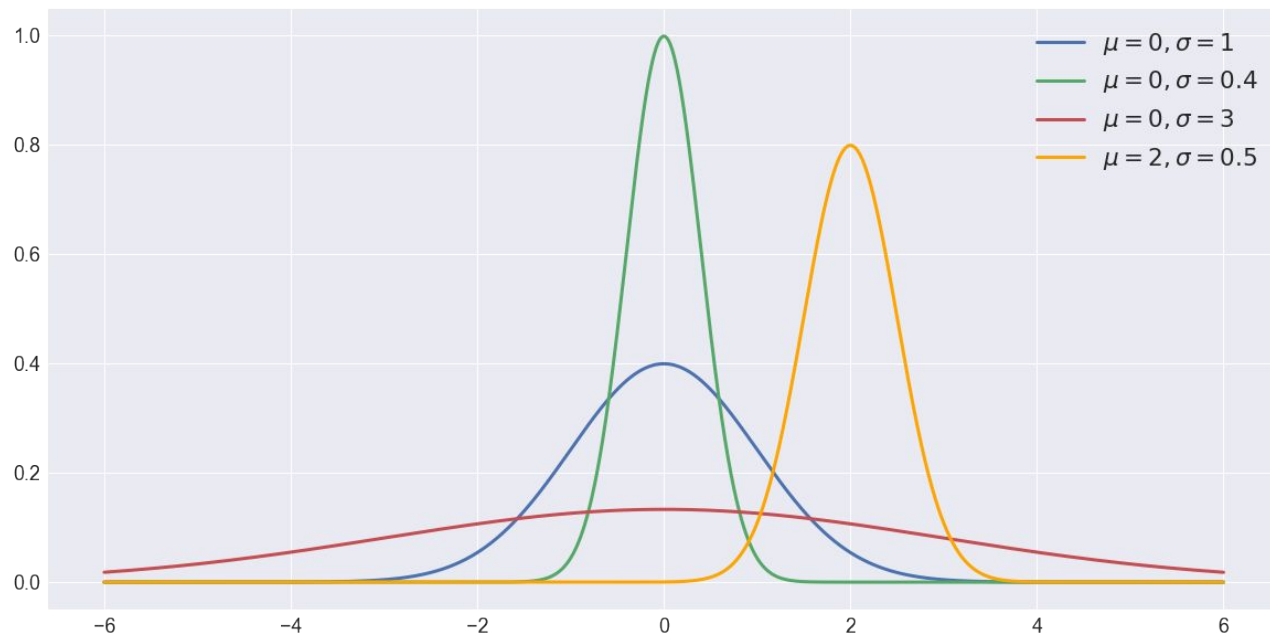


# the Normal distribution

- also known as a Gaussian / bell curve
- Normal distributions are symmetric around their mean
- The mean, median, and mode of a normal distribution are equal
- The area under the bell curve is equal to 1.0
- Normal distributions are denser in the center and less dense in the tails
- Normal distributions are defined by two parameters, the mean and the standard deviation
- Around 68% of the area of a normal distribution is within one standard deviation of the mean
- Approximately 95% of the area of a normal distribution is within two standard deviations of the mean



# expected value & variance

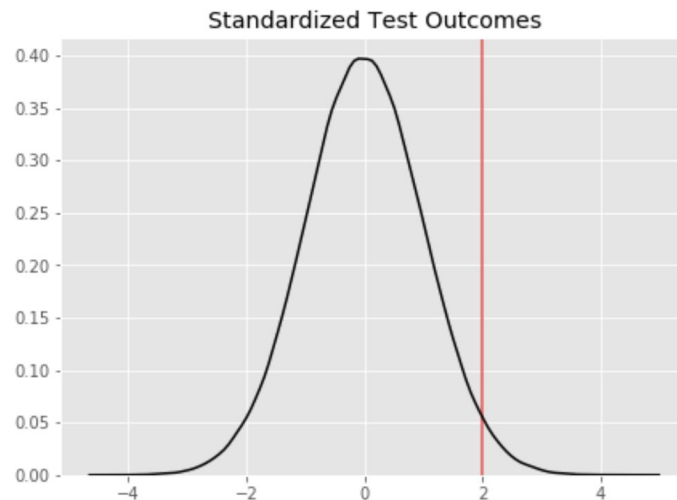


- $E(X) = \mu$
- $\text{Var}(X) = \sigma^2$

PDF	$\frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$
CDF	$\frac{1}{2} \left[ 1 + \text{erf}\left(\frac{x-\mu}{\sigma\sqrt{2}}\right) \right]$

# the Standard Normal distribution

- $E(X) = 0$ ,  $\text{Var}(X) = 1$
- Allows us to compare different normal distributions
- z-scores!



# z-score

- a z-score tells us how many standard deviations away from the mean a point would be in a Standard Normal distribution
- z-scores are associated with cumulative probabilities, retrieved from the [z-table](#)
- $z = (x - \mu) / \sigma$  for a single point
- $z = (x - \mu) / (\sigma / \sqrt{n})$  for a sample

