

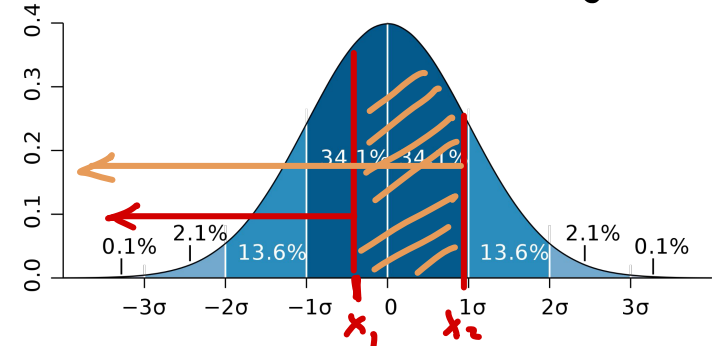
ME 103 Discussion 2

Week of 1/26

Probabilities and Normal Distribution

$n > 30 \rightarrow z$
 $n < 30 \rightarrow t$

- General Concept: Taking your values, standardizing them to a normal distribution (from 0-1) and using mean (μ), std.dev(σ) and z table to find probabilities
- Z-score: $n > 30$, t-score $n < 30$ UNLESS distribution is assumed to be normal
- Estimate of population mean = sample mean (\bar{x}) $\rightarrow np.mean()$
- Estimate of population standard deviation = sample standard deviation (s) $\rightarrow np.std(), ddof=1$
- Converting to z-score:
 - $Z = (X - \mu) / \sigma$
- Plug into z-table (first two digits on columns, last digit in first row), or norm.cdf in python
- Cumulative Density Function (CDF) value =
 - I.g. $\Phi(1.96) = 95\% \rightarrow 95\%$ of the distribution exists at 1.96 and below.



z	0	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
-0	.5000	.4960	.4920	.4880	.4840	.4800	.4760	.4720	.4681	.4641
0.1	.4601	.4562	.4522	.4482	.4443	.4403	.4364	.4325	.4285	.4245
0.2	.4207	.4168	.4128	.4088	.4049	.4010	.3970	.3931	.3891	.3851
0.3	.3810	.3770	.3730	.3691	.3651	.3611	.3571	.3531	.3491	.3451
0.4	.3410	.3370	.3330	.3291	.3251	.3211	.3171	.3131	.3091	.3051
0.5	.3010	.2970	.2930	.2891	.2851	.2811	.2771	.2731	.2691	.2651
0.6	.2610	.2570	.2530	.2491	.2451	.2411	.2371	.2331	.2291	.2251
0.7	.2210	.2170	.2130	.2091	.2051	.2011	.1971	.1931	.1891	.1851
0.8	.1810	.1770	.1730	.1691	.1651	.1611	.1571	.1531	.1491	.1451
0.9	.1410	.1370	.1330	.1291	.1251	.1211	.1171	.1131	.1091	.1051
1	.1010	.0970	.0930	.0891	.0851	.0811	.0771	.0731	.0691	.0651
1.1	.0610	.0570	.0530	.0491	.0451	.0411	.0371	.0331	.0291	.0251
1.2	.0210	.0170	.0130	.0091	.0051	.0011	.0000	.0000	.0000	.0000
1.3	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
1.4	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
1.5	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
1.6	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
1.7	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
1.8	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
1.9	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
2	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
2.1	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
2.2	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
2.3	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
2.4	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
2.5	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
2.6	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
2.7	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
2.8	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
2.9	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
3	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
3.1	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
3.2	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
3.3	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
3.4	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
3.5	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
3.6	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
3.7	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
3.8	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
3.9	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000
4	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000	.0000

$\mu = 6$
 $CI = \pm 0.2$
 $\alpha = 0.95$
 $CI = \pm Z\sigma$
 $0.2 = \pm 1.96\sigma$
 $P(X > 6.050) = 1 - P(X < 6.050) = 1 - \Phi\left(\frac{x - \mu}{\sigma}\right)$

Basic Probability Rules!

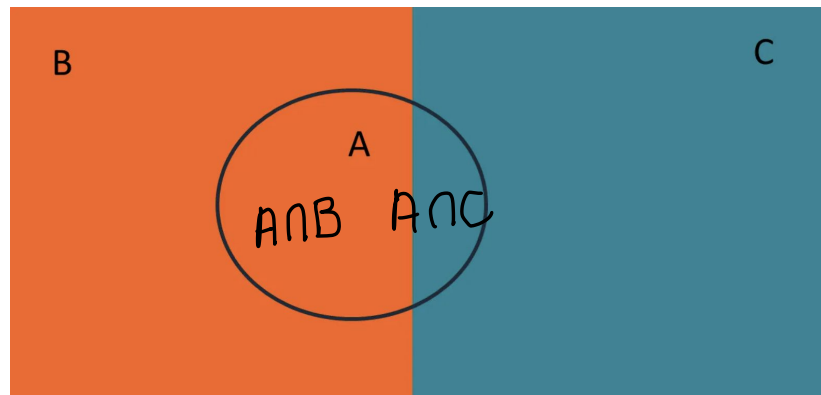
1. Possible values for probabilities range from 0 to 1
 - a. 0 = impossible event, 1 = certain event
2. The sum of all the probabilities for all possible outcomes is equal to 1.
 - a. $P(\Omega) = 1$, where Ω is the probability space- set of all possible events.
3. Complement Rule (Not A):
 - a. $P(\bar{A}) = P(A^c) = 1 - P(A)$
4. Union Rule (A or B):
 - a. $P(\underline{A} \cup B) = P(A) + P(B) - P(A \cap B)$
↓ intersection
5. Product Rule (A and B):
 - a. $P(A \cap B) = P(A) * P(B)$ ***for independent events
6. Conditional Probability (A *given* B):
 - a. $P(A|B) = P(A \cap B) / P(B)$

Total Probability Rule

- By using conditional probabilities, we can find the probability of unknown event A.
- Ex. (right)
 - $P(A) = P(A \cap B) + P(A \cap C)$
- We also know: $P(A \cap B) = P(A|B) * P(B)$ (Conditional Rule)
- This can be summed through the nth conditional probability.

$$\Pr(B) = \Pr(A_1 \cap B) + \dots + \Pr(A_n \cap B) = \sum_{i=1}^n \Pr(A_i) \Pr(B|A_i)$$

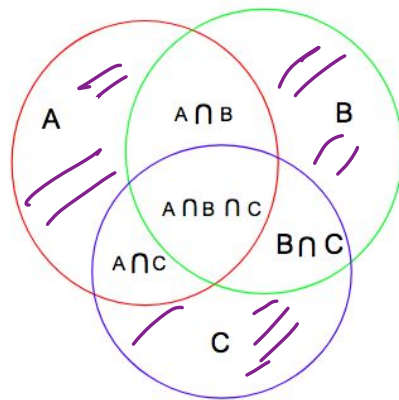
$$P(B) = \sum P_r(A)P(B|A) + P(C)P(B|C)$$



Inclusion/ Exclusion Principle

- Extension of Union Law.
- Math version \rightarrow

$$\Pr\left(\bigcup_{i=1}^n A_i\right) = \sum \Pr(A_i) - \sum_{i<j} \Pr(A_i \cap A_j) + \sum_{i<j<k} \Pr(A_i \cap A_j \cap A_k) - \dots$$



$$|A \cup B \cup C| = |A| + |B| + |C| - |A \cap B| - |A \cap C| - |B \cap C| + |A \cap B \cap C|.$$

Bayes Theorem

- Rearrangement of previous rules

Conditional probability: Bayes' Theorem

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

Probability Potpourri Part a I

- Split into random vars:
 - $P(W)$: probability windy = 0.3
 - $P(H)$: probability hit
 - $P(H|W) = 0.4$: "archer hits her target given it is windy
- Find probability it is windy and she hits the shot = $P(H \cap W)$
 - Solve using probability rules and knowns
- Find the probability she hits on her first shot.
 - You have two scenarios, that she hits and it is windy + probability she hits and its not windy.
 - Whenever you have the probability of an event over multiple scenarios, use total probability rule.

$$P(W) = 0.3 \quad P(\bar{W}) = 0.7 \quad P(H|\bar{W}) = 0.7$$

$$P(H) = \quad P(\bar{H}) =$$

$$P(H|W) = 0.4 \quad P(H \cap W) = \checkmark$$

$$P(\bar{H}|W) = 1 - P(H|W) = 0.6$$

or equivalently,

$$P(A) = P(A|B_1)P(B_1) + P(A|B_2)P(B_2) + \dots + P(A|B_k)P(B_k).$$



^^ total probability rule

$$B_1 = W \quad B_2 = \bar{W}$$

$$P(H) = P(H|W)P(W) + P(H|\bar{W})P(\bar{W})$$

Probability Potpourri II



Binomial Distribution Formula



- Find the probability she hits **once** in **two shots**:
 - Number of trials: $n = 2$
 - Number of successes: $x = 1$
 - Probability of success (she hits once) = p calculated from previous question
- Use the binomial distribution formula

$$P(\bar{H}) = 1 - P(H) \longrightarrow$$

$$P(x) = \binom{n}{x} p^x q^{n-x} = \frac{n!}{(n-x)!x!} p^x q^{n-x}$$

where

n = the number of trials (or the number being sampled)

x = the number of successes desired

p = probability of getting a success in one trial

$q = 1 - p$ = the probability of getting a failure in one trial

Probability Potpourri III

- “on an occasion when she missed, there was no gust of wind.”
 - No wind, **given** no hit
 - You have every probability now, can use Bayes

$$P(\bar{W}|\bar{H}) = \frac{P(\bar{H}|\bar{W})P(\bar{W})}{P(\bar{H})}$$

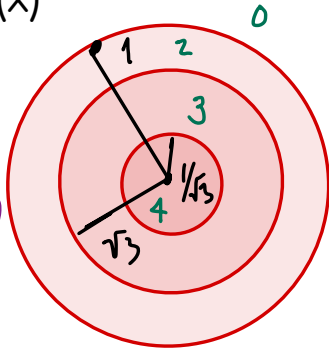
$$P(A|B) = \frac{P(B|A) P(A)}{P(B)}$$

$$f(x) = \frac{2}{\pi(1+x^2)}$$

$$E[X] = \int_0^{1/\sqrt{3}} 4 f(x) dx + \int_{1/\sqrt{3}}^{\sqrt{3}} 3 f(x) dx + \dots$$

Expected Value

- X = point values, 4, 3, 2
- Plug into matlab or solve by hand
- Separate into 3 limits $(0, 1/\sqrt{3})$, $(1/\sqrt{3}, \sqrt{3})$, $(\sqrt{3}, 1)$
- Helpful rule: $\int (1)/(1+x^2) = \arctan(x)$



$$\int \frac{1}{1+x^2} = \arctan(x)$$

$$E[X] = \sum_i x_i f(x_i)$$

↳ discrete

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

↳ continuous

Shock Dyno

$X \rightarrow \{1, \dots, 6\}$
 $\uparrow 100x$

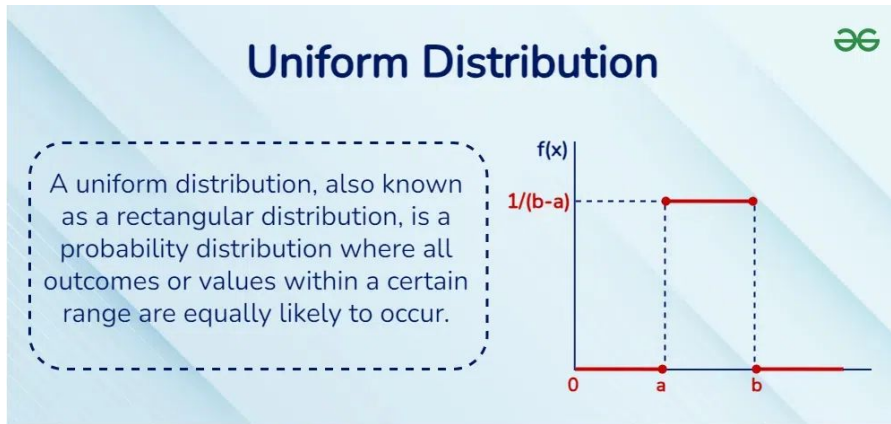
\rightarrow discrete Uniform

$Y \rightarrow 1 \text{ or } 0$
 $P(Y) = 0.5$

Binomial

$\leftarrow 600x$

- 2 random variables, X and Y. \swarrow
- You're looking for $P(X < Y)$, or $P(X - Y) < 0$.
Just like before, need to convert to z scores and find CDF
- Pick out "equal likelihood per stroke" gives you a hint that you are using uniform distribution for X.
- $\{1, \dots, 6\}$ hints that it is a discrete uniform distribution.
 - Find $E[X]$: $n = 6$,
 - Find var of X
- Since you have 100 independent samples, **The total variance = the variance of 1 trial x 100.**



Discrete Uniform: $X \sim \text{Uniform}(n)$

$$p_x = \frac{1}{n} \quad (x = 1, 2, \dots, n) \quad \mu = (n + 1)/2, \quad \sigma^2 = (n^2 - 1)/12$$

\uparrow
 $E[X]$

\uparrow
 $\text{Var}[\sigma^2]$

Shock Dyno

- For Y: “Each window independently records 1 (event) or 0 (no event) with probability 0.5 each.”
 - Whenever you have a chance of 1 or 0 only, success/fail, that’s a hint that you use Binomial distribution.
 - $n = 600$
 - $p = 0.5$
 - Find mean and variance, remember var * 600 trials = total variance.
- Define random variable $Z = X - Y$
- We can assume normal distribution.
- Find $E[Z] = E[X - Y]$ and $\text{Var}[Z] = \text{Var}[X - Y]$.
 $\text{Cov} = 0$ for independent variables.
- Finally, convert 0 to z score and use CDF to find percent of distribution under $(0 - E[Z]) / \sqrt{\text{Var}[Z]}$

Binomial distribution equations

$$\text{Mean} = \mu = E(x) = np \quad \leftarrow E[Y]$$

$$\text{Variance} = \sigma^2 = np(1 - p) \quad \leftarrow \text{Var}[Y]$$

$$\text{Standard Deviation} = \sigma = \sqrt{np(1 - p)}$$

where

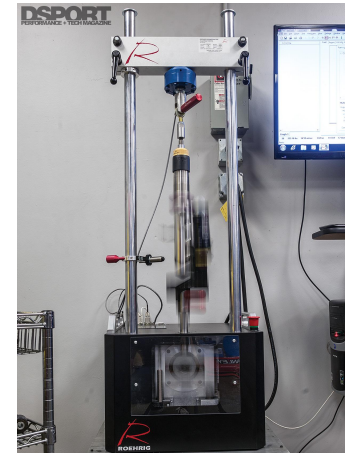
$$n = \text{number of trials} \quad E[x] = 100 \cdot E[x_i]$$

$$p = \text{probability of success} \quad \text{Var}[x] = 100 \cdot \text{Var}[x_i]$$

$$1 - p = \text{probability of other outcome (failure)}$$

$\mathbb{E}[\cdot]$ is a linear operator: $E[X + Y] = E[X] + E[Y]$

$$\text{Var}(X - Y) = \text{Var}(X) + \text{Var}(Y) - 2 \text{Cov}(X, Y)$$



Larry Time



- (d) [6 pts] You meet two students in Hesse Hall. Assume that each student is a Senior or a Sophomore with equal probability, and each student takes ME103 with probability $1/10$, independent of each other and independent of their class standing. What is the probability that both students are Seniors, given at least one of them is a Senior currently taking ME103?

- (e) [6 pts] There are n identical looking sensors in Hesse 122 and n boxes to store them, where sensor i is supposed to go into box i . After a hectic cleanup, the sensors get thrown into the boxes uniformly at random (every permutation is equally likely). Assume n is a positive integer. What is the probability that **no sensor** ends up in its correct bin? Also find the limit of this probability as $n \rightarrow \infty$.

Hint: Use the principle of inclusion-exclusion and the power series representation of e^x . Assume that each box gets exactly one sensor.

