Probability

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

If A and B are independent $P(A \cap B) = P(A)P(B)$

Expected Value

$$E[x] = \int_{-\infty}^{\infty} x p(x) dx \qquad E[g(x)] = \int_{-\infty}^{\infty} g(x) p(x) dx$$

Variance

$$Var[x] = E[(x - E[x])^{2}] = \sigma^{2} = \int_{-\infty}^{\infty} (x - E[x])^{2} p(x) dx$$
$$E((x - E[x])^{2}) = E[x^{2}] - (E[x])^{2}$$

Covariance

$$cov(X_1, X_2) = E[(X_1 - m_1)(X_2 - m_2)]$$

= $E[(X_1)(X_2)] - m_1 m_2$

Correlation

$$\rho = \frac{cov(X_1, X_2)}{\sqrt{V(X_1)V(X_2)}}$$

Probability Distribution

Normal

$$\frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} \qquad E[X] = \mu \qquad Var[X] = \sigma^2$$

Exponential

$$\lambda e^{-\lambda x} \qquad E[X] = \frac{1}{\lambda} \qquad Var[X] = \frac{1}{\lambda^2}$$

Uniform

$$\begin{cases} \frac{1}{b-a} & \text{for } x \in [a,b] \\ 0 & \text{otherwise} \end{cases}$$

$$E[X] = \frac{1}{2}(a+b) \qquad Var[X] = \frac{1}{12}(b-a)^2$$

Bernoulli

$$\begin{cases} q = 1 - p & \text{if } k = 0 \\ p & \text{if } k = 1 \end{cases} \quad E[X] = p \quad Var[X] = p(1 - p)$$

Binomial

$$\binom{n}{k}p^k(1-p)^{n-k} \qquad E[X] = np \qquad Var[X] = np(1-p) \qquad n = \frac{m+1}{m}\left(\frac{(Z_\alpha + Z_\beta)\sigma}{MDE}\right)^2$$

Poisson

$$\frac{\lambda^k e^{-\lambda}}{k!}$$
 $E[X] = \lambda$ $Var[X] = \lambda$

Pareto

$$\frac{\alpha x_m^a}{x^{\alpha+1}} \qquad E[x] = \begin{cases} \infty & \text{for } \alpha \le 1\\ \frac{\alpha x_m}{\alpha - 1} & \text{for } \alpha > 1 \end{cases}$$

$$Var[X] = \begin{cases} \infty & \text{for } \alpha \le 1\\ \frac{x_m^2 \alpha}{(\alpha - 1)^2 (\alpha - 2)} & \text{for } \alpha > 1 \end{cases}$$

MLE

To find the MLE given data

- 1. The likelihood function $P(data|\lambda)$, λ is the parameter
- 2. $\frac{d}{d\lambda}(\log \text{ likelihood}) = 0$, Find λ

Hypothesis Testing

Errors

		True State of Nature	
		H_0	H_A
Our Decision	Reject H_0	Type-I Error	correct decision
	Reject H_0	correct decision	Type-II Error

One Sample z-test

Use when the **Variance** (σ^2) of the data is known

$$z = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}}$$

Two Sample z-test

One Sample t-test

$$t = \frac{\bar{x} - \mu}{s / \sqrt{n}}$$
 where $s^2 = \frac{1}{n - 1} \sum_{i=1}^{n} (x_i - \bar{x})^2$ $df = n - 1$

Two Sample t-test with Equal Variance

Assume that the data have $\sigma_1 = \sigma_2$

Two Sample t-test with Unequal Variance

Assume that the data have $\sigma_1 \neq \sigma_2$

A/B Testing

MDE

Sample Size

m is the split rate (50:50 = 1, 80:20 = 4)

$$n = \frac{m+1}{m} \left(\frac{(Z_{\alpha} + Z_{\beta})\sigma}{MDE} \right)^{2}$$

When not to do an A/B test

- \bullet Things that cannot be summarized into one or a few metrics
- Totally new things

- Delayed results
- One-off events
- Cannot split group independently