Stats Formula Sheet

Mean:
$$\overline{y} = \frac{1}{n} \sum_{i=1}^{n} y_i$$

Variance:
$$s^2 = \frac{1}{n-1} \sum_{i=1}^{n} (y_i - \overline{y})^2$$

Standard Deviation:
$$s = \sqrt{s^2}$$

Permutation Equation:
$$P_r^n = \frac{n!}{(n-r)!}$$

Combination Equation:
$$\binom{n}{r} = C_r^n = \frac{P_r^n}{r!} = \frac{n!}{r!(n-r)!}$$

Conditional Probability:
$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

Independent and Dependent:

Events A and B are independent if:

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

$$OR P(A \cap B) = P(A)P(B)$$

Otherwise, events are dependent

If events are dependent,

$$P(A \cap B) = P(A)P(B|A) = P(B)P(A|B)$$

If events are independent,

$$P(A \cap B) = P(A)P(B)$$

If events are mutually exclusive,

$$P(A \cap B) = 0$$
 AND

$$P(A \cup B) = P(A) + P(B)$$

Otherwise,

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

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Complement: $P(A) = 1 - P(\overline{A})$

Theorem of Total Probability: $P(A) = \sum_{i=1}^{n} P(A|B_i)P(B_i)$

Bayes' Rule:
$$P(B_j|A) = \frac{P(A \cap B_j)}{P(A)} = \frac{P(A|B_j)P(B_j)}{\sum_{i=1}^k P(A|B_i)P(B_i)}$$

When Y is a discrete random variable,

Expected Value: $E(Y) = \sum_{y} yp(y)$

Variance: $V(Y) = E[(y - E(Y))^2]$

Standard Deviation: $S(Y) = \sqrt{V(Y)}$

Binomial Distributions: $p(y) = \binom{n}{y} p^y q^{n-y}$

Expected Value: E(Y) = np

Variance: V(Y) = npq.

Standard Deviation: \sqrt{npq}

Geometric Distributions: $p(y) = q^{y-1}p$

Expected Value: E(Y) = 1/p

Variance: $V(Y) = \frac{1-p}{p^2}$

Standard Deviation: $\sqrt{\frac{1-p}{p^2}}$

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Hypergeometric Distributions:
$$p(y) = \frac{\binom{r}{y} \times \binom{N-r}{n-y}}{\binom{N}{n}}$$

Expected Value:
$$E(Y) = \frac{nr}{N}$$

Variance:
$$V(Y) = n(\frac{r}{n})(\frac{N-r}{N})(\frac{N-r}{N-1})$$

Standard Deviation:
$$S(Y) = \sqrt{n(\frac{r}{n})(\frac{N-r}{N})(\frac{N-n}{N-1})}$$

Poisson Distributions:
$$p(y) = (\frac{\lambda^y}{y!}) \times e^{-\lambda}$$

Expected Value:
$$E(Y) = \lambda$$

Variance:
$$V(Y) = \lambda$$

Standard Deviation:
$$S(Y) = \sqrt{\lambda}$$

Chebyshev's Theorem

$$P(|Y - \mu| < k\sigma) \ge 1 - \frac{1}{k^2}$$
 OR $P(|Y - \mu| \ge k\sigma) \le \frac{1}{k^2}$

- Lower Interval Bound = $\mu k\sigma$
- Upper Interval Bound = $\mu + k\sigma$