

# AML5103: Applied Probability & Statistics – Formula Sheet – August 14, 2023

## 1. Counting formulas:

Setup	Formula
# of ways to select ( <i>without replacement</i> ) and arrange ( <i>order matters</i> ) $n$ distinct objects # of ways to distribute $n$ distinct objects into $n$ distinct bins with one object per bin	$n!$
# of ways to select ( <i>without replacement, order doesn't matter</i> ) $r$ objects out of $n$ distinct objects # of ways to arrange $n$ objects when one object repeats $r$ times and the other $n - r$ times # of ways to divide $n$ distinct objects into two unlabeled groups of unequal sizes $r$ and $n - r$ # of ways to divide $n$ distinct objects into two labeled groups of specific sizes $r$ and $n - r$	$\binom{n}{r} = nC_r$
# of ways to select ( <i>without replacement</i> ) and arrange ( <i>order matters</i> ) $r$ objects out of $n$ distinct objects # of ways to distribute $r$ distinct objects into $n$ distinct bins with at most one object per bin	$r! \binom{n}{r} = nP_r$
# of ways to select ( <i>with replacement</i> ) and arrange ( <i>order matters</i> ) $r$ objects out of $n$ distinct objects # of ways to distribute $r$ distinct objects into $n$ distinct bins	$n^r$
# of ways to arrange $n$ objects with $r_1, r_2, \dots, r_k$ repetitions, where $n = r_1 + r_2 + \dots + r_k$ # of ways to divide $n$ distinct objects into $k$ unlabeled groups of unequal sizes $r_1, r_2, \dots, r_k$ , where $n = r_1 + r_2 + \dots + r_k$ # of ways to divide $n$ distinct objects into $k$ labeled groups of specific sizes $r_1, r_2, \dots, r_k$ , where $n = r_1 + r_2 + \dots + r_k$	$\frac{n!}{r_1! r_2! \dots r_k!}$
# of ways to divide $n$ distinct objects into $k$ unlabeled groups with some of equal sizes, say $\underbrace{r_1 = r_2}_{2 \text{ groups}}, \underbrace{r_3 = r_4 = r_5}_{3 \text{ groups}}$ , and $r_6, \dots, r_k$ are different such that $n = r_1 + r_2 + \dots + r_k$	$\left( \frac{1}{2!3!} \right) \left( \frac{n!}{r_1! r_2! \dots r_k!} \right)$
# of ways to select $r$ objects ( <i>with replacement, order doesn't matter</i> ) out of $n$ distinct objects # of ways to distribute $r$ identical objects into $n$ distinct bins # of non-negative integer solutions to the equation $x_1 + x_2 + \dots + x_n = r$	$\binom{n+r-1}{r}$
# of ways to distribute $r$ identical objects into $n$ distinct bins such that no bin is empty # of positive integer solutions to the equation $x_1 + x_2 + \dots + x_n = r$	$\binom{r-1}{n-1}$

2. **Inclusion–exclusion principle for 4 events:**

$$\begin{aligned}
 P(A_1 \cup A_2 \cup A_3 \cup A_4) = & P(A_1) + P(A_2) + P(A_3) + P(A_4) \\
 & - P(A_1 \cap A_2) - P(A_1 \cap A_3) - P(A_1 \cap A_4) - P(A_2 \cap A_3) - P(A_2 \cap A_4) - P(A_3 \cap A_4) \\
 & + P(A_1 \cap A_2 \cap A_3) + P(A_1 \cap A_2 \cap A_4) + P(A_1 \cap A_3 \cap A_4) + P(A_2 \cap A_3 \cap A_4) \\
 & - P(A_1 \cap A_2 \cap A_3 \cap A_4).
 \end{aligned}$$

3. **Conditional probability:**

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \Leftrightarrow P(A \cap B) = P(A|B) \times P(B).$$

4. **Law of total probability:**

For event  $A$  and the mutually exclusive & collectively exhaustive set of events  $\{B_1, B_2, \dots, B_k\}$  for the same sample space,

$$\begin{aligned}
 P\left(\underbrace{A}_{\text{e.g., has Cancer}}\right) &= P\left(\underbrace{A \cap B_1}_{\text{has Cancer AND from State-1}}\right) + P\left(\underbrace{A \cap B_2}_{\text{has Cancer AND from State-2}}\right) + \dots + P\left(\underbrace{A \cap B_k}_{\text{has Cancer AND from State-k}}\right) \\
 &= P(A|B_1) \times P(B_1) + P(A|B_2) \times P(B_2) + \dots + P(A|B_k) \times P(B_k).
 \end{aligned}$$

**Additional conditioning form:**

$$\begin{aligned}
 P\left(\underbrace{A}_{\text{e.g., has Cancer}} \mid \text{male}\right) &= P\left(\underbrace{A \cap B_1}_{\text{has Cancer AND from State-1}} \mid \text{male}\right) + P\left(\underbrace{A \cap B_2}_{\text{has Cancer AND from State-2}} \mid \text{male}\right) + \dots + P\left(\underbrace{A \cap B_k}_{\text{has Cancer AND from State-k}} \mid \text{male}\right) \\
 &= P(A|B_1, \text{male}) \times P(B_1 | \text{male}) + P(A|B_2, \text{male}) \times P(B_2 | \text{male}) + \dots + P(A|B_k, \text{male}) \times P(B_k | \text{male}).
 \end{aligned}$$

5. **Bayes' theorem:**

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(B \cap A)}{P(B)} = \frac{P(B|A) \times P(A)}{P(B)} = \frac{P(B|A) \times P(A)}{P(B|A) \times P(A) + P(B|A^c) \times P(A^c)}$$

**Additional conditioning form:**

$$P(A|B, C) = \frac{P(B|A, C) \times P(A|C)}{P(B|C)} = \frac{P(B|A, C) \times P(A|C)}{P(B|A, C) \times P(A|C) + P(B|A^c, C) \times P(A^c|C)}.$$

## 6. Discrete random variables:

Random Variable $X$	Parameters	Values Taken $j$ or $x$	PMF $P_X(j)$ $= P(X = j)$	Expected Value $E[X]$ $= \sum_j j P_X(j)$	Variance $Var[X]$ $= \sum_j (j - E[X])^2 P_X(j)$
Bernoulli	$p$	$j = 0, 1$	$p^j(1-p)^{1-j}$	$p$	$p(1-p)$
Multinoulli	$p_1, p_2, \dots, p_k$	$x = \underbrace{\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}}_{\text{assuming } k=4}$	$\prod_{i=1}^k p_i^{I(x_i=1)}$	$\underbrace{\begin{bmatrix} p_1 \\ p_2 \\ p_3 \\ p_4 \end{bmatrix}}_{\text{assuming } k=4}$	$\underbrace{\begin{bmatrix} p_1(1-p_1) \\ p_2(1-p_2) \\ p_3(1-p_3) \\ p_4(1-p_4) \end{bmatrix}}_{\text{assuming } k=4}$
Binomial	$n, p$	$j = 0, 1, \dots, n$	$\binom{n}{j} p^j (1-p)^{n-j}$	$np$	$np(1-p)$
Multinomial	$n, p_1, p_2, \dots, p_k$	$x = \underbrace{\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}}_{\text{assuming } k=4}, \text{ where } \sum_{i=1}^4 x_i = n$	$\underbrace{\binom{n}{x_1, x_2, x_3, x_4} \prod_{i=1}^4 p_i^{x_i}}_{\text{assuming } k=4}$	$\underbrace{\begin{bmatrix} np_1 \\ np_2 \\ np_3 \\ np_4 \end{bmatrix}}_{\text{assuming } k=4}$	$\underbrace{\begin{bmatrix} np_1(1-p_1) \\ np_2(1-p_2) \\ np_3(1-p_3) \\ np_4(1-p_4) \end{bmatrix}}_{\text{assuming } k=4}$
Negative Binomial	$r, p$	$j = r, r+1, \dots$	$\binom{j-1}{r-1} p^r (1-p)^{j-r}$	$\frac{r}{p}$	$\frac{r(1-p)}{p^2}$
Geometric	$p$	$j = 1, 2, \dots$	$(1-p)^{j-1} p$	$\frac{1}{p}$	$\frac{1-p}{p^2}$
Poisson	$\lambda$	$j = 0, 1, \dots$	$\frac{e^{-\lambda} \lambda^j}{j!}$	$\lambda$	$\lambda$
Hypergeometric	$n_{\text{success}}, n_{\text{failure}}, n$	$j = 0, 1, \dots, n$	$\frac{\binom{n_{\text{success}}}{j} \binom{n_{\text{failure}}}{n-j}}{\binom{n_{\text{success}}+n_{\text{failure}}}{n}}$	where $p = \frac{np}{n_{\text{success}}+n_{\text{failure}}}$	where $N = n_{\text{success}} + n_{\text{failure}}$ $\frac{np(1-p)(N-n)}{N-1}$

## 7. Connections:

- $X \sim \text{Bin}(n=1, p)$  is same as  $X \sim \text{Bernoulli}(p)$ .
- $X \sim \text{NegBin}(r=1, p)$  is same as  $X \sim \text{Geom}(p)$ .
- If  $X \sim \text{HypGeom}(n_{\text{success}}, n_{\text{failure}}, n)$  such that  $n \ll (n_{\text{success}} + n_{\text{failure}})$  and  $n_{\text{success}}$  is not close to 0 and  $n_{\text{success}} + n_{\text{failure}}$  then  $X \sim \text{Bin}(n, p)$  approximately where  $p = n_{\text{success}} / (n_{\text{success}} + n_{\text{failure}})$ .
- For  $n \gg p$ ,  $X \sim \text{Bin}(n, p)$  implies  $X \sim \text{Poi}(\lambda = n \times p)$  approximately.