

# STA9715

## Test 1 - Formula Sheet

### Foundations:

- Probabilities are Non-Negative:  $0 \leq \mathbb{P}(A) \leq 1$  for all *events*  $A$
- Probability of Entire Sample Space:  $\mathbb{P}(\Omega) = 1$
- Probability of Empty Set:  $\mathbb{P}(\emptyset) = 0$
- 'Union Bound':  $\mathbb{P}(A \cup B) \leq \mathbb{P}(A) + \mathbb{P}(B)$  with equality if  $A, B$  are *disjoint* ( $A \cap B = \emptyset$ )
- Complements:  $\mathbb{P}(A^c) = 1 - \mathbb{P}(A)$
- Naive Probability:  $\mathbb{P}(A) = |A|/|\Omega|$
- Counting:  ${}_nP_k = \frac{n!}{(n-k)!}$  (permutations - *ordered* choice of  $k$  from  $n$ );  ${}_nC_k = \binom{n}{k} = \frac{n!}{(n-k)!k!}$  (combinations - *unordered*)
- DeMorgan's Laws:  $(A \cup B)^c = A^c \cap B^c$  and  $(A \cap B)^c = A^c \cup B^c$

### Random Variables (Discrete):

- $\mathbb{P}(X = x) = f_X(x)$  is the *probability mass function* of  $X$ . Gives the probability of observing  $X = x$  exactly
- $\sum_{x \in \text{supp}(X)} \mathbb{P}(X = x) = 1$ .  $0 \leq f_X(x) \leq 1$

### Random Variables (Continuous):

- *Probability density function* of  $X$ :  $f_X(x)$  Integrates to give the probability  $X$  in an interval:  $P(a \leq X \leq b) = \int_a^b f_X(x) dx$
- $\int_{x \in \text{supp}(X)} f_X(x) dx = 1$ .  $f_X(x)$  may be greater than 1 for small ranges; never negative

### Moments:

- Expectation:  $\mathbb{E}[X] = \sum x * f_X(x)$  or  $\int x * f_X(x) dx$
- Linearity of Expectation:  $\mathbb{E}[aX + bY + c] = a\mathbb{E}[X] + b\mathbb{E}[Y] + c$
- Expectation of Functions:  $\mathbb{E}[g(X)] = \sum g(x)f_X(x)$  or  $\int g(x)f_X(x) dx$
- Variance:  $\mathbb{V}[X] = \text{Var}(X) = \mathbb{E}[(X - \mathbb{E}[X])^2] = \mathbb{E}[X^2] - \mathbb{E}[X]^2 \geq 0$
- $\mathbb{V}[aX + bY + c] = a^2\mathbb{V}[X] + b^2\mathbb{V}[Y]$  only if  $X, Y$  are uncorrelated. (Independent implies uncorrelated)
- Expectation of Indicators:  $\mathbb{E}[1_{\in A}(X)] = \mathbb{P}(X \in A)$ . Useful to reduce probabilities to linear expectation calculations

### Conditional Probabilities:

- $\mathbb{P}(A|B) = \mathbb{P}(A \cap B)/\mathbb{P}(B)$ . Special case:  $A \subseteq B \implies \mathbb{P}(A|B) = \mathbb{P}(A)/\mathbb{P}(B)$
- Bayes' Rule:  $\mathbb{P}(A|B) = \mathbb{P}(B|A) * \mathbb{P}(A)/\mathbb{P}(B) = \mathbb{P}(B|A) * \mathbb{P}(A)/(\mathbb{P}(B|A) * \mathbb{P}(A) + \mathbb{P}(B|A^c) * \mathbb{P}(A))$
- Law of Total Probability: if  $\{A_j\}$  are a disjoint partition of  $\Omega$  then  $\mathbb{P}(B) = \sum_j \mathbb{P}(B|A_j)\mathbb{P}(A_j)$
- Law of Total Expectation:  $\mathbb{E}[X] = \sum \mathbb{E}[X|A_i]\mathbb{P}(A_i)$  for partition  $\{A_i\}$  or  $\mathbb{E}_X[X] = \mathbb{E}_Y[\mathbb{E}_X[X|Y]]$
- Law of Total Variance:  $\mathbb{V}[X] = \mathbb{E}_Y[\mathbb{V}_X[X|Y]] + \mathbb{V}_Y[\mathbb{E}_X[X|Y]]$
- Independence:  $A, B$  are *independent* if  $\mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B)$ . Equivalently  $\mathbb{P}(B|A) = \mathbb{P}(B)$  and  $\mathbb{P}(A|B) = \mathbb{P}(A)$

### Distributions:

- Bernoulli:  $X \sim \text{Bern}(p)$ .  $\mathbb{P}(X = 1) = p$ ;  $\mathbb{P}(X = 0) = 1 - p$ .  $\mathbb{E}[X] = p$ ;  $\mathbb{V}[X] = p(1 - p)$ . 'Coin Flip'
- Binomial: Sum of  $n$  IID Bernoulli:  $P(X = x) = \binom{n}{x}p^x(1 - p)^{n-x}$ .  $\mathbb{E}[X] = np$ .  $\mathbb{V}[X] = np(1 - p)$
- Poisson: Limit of  $n \rightarrow \infty, p \rightarrow 0$  Binomial.  $X \sim \text{Pois}(\mu)$ .  $\mathbb{P}(X = x) = \mu^x e^{-\mu} / x!$ .  $\mathbb{E}[X] = \mathbb{V}[X] = \mu$
- Geometric: IID Bernoulli 'until' 1st success:  $X \sim \text{Geom}(p)$ .  $\mathbb{P}(X = x) = p(1 - p)^{x-1}$ .  $\mathbb{E}[X] = 1/p$ .  $\mathbb{V}[X] = (1 - p)/p^2$ . Memoryless
- Normal:  $X \sim \mathcal{N}(\mu, \sigma^2)$ .  $f_X(x) = e^{-(x-\mu)^2/2\sigma^2} / \sqrt{2\pi\sigma^2}$ .  $\mathbb{E}[X] = \mu$ .  $\mathbb{V}[X] = \sigma^2$ . 'Bell curve'
- Exponential:  $X \sim \text{Exp}(\lambda)$ .  $f_X(x) = \lambda e^{-\lambda x}$ .  $\mathbb{E}[X] = \lambda^{-1}$ .  $\mathbb{V}[X] = \lambda^{-2}$ . Continuous geometric
- Uniform (Discrete and Continuous).  $\mathbb{E}[\text{DUnif}\{a, \dots, b\}] = \mathbb{E}[\text{CUnif}([a, b])] = (a + b)/2$ .  $\mathbb{V}[\text{CUnif}([a, b])] = (b - a)^2/12$