

# Formula Sheet – MATH 401 – Final Exam – Spring2022

1. **Sample mean:**  $\bar{X} = \frac{\sum X}{n}$ , & **Sample variance:**  $s^2 = \frac{\sum X^2 - \frac{(\sum X)^2}{n}}{n-1}$ .
2. **Binomial:**  $P(X = k) = {}^nC_k \cdot p^k \cdot (1-p)^{n-k}$ ,  $E(X) = np$ ,  $V(X) = np(1-p)$ .
3. **Geometric:**  $P(X = k) = p \cdot (1-p)^{k-1}$ ,  $E(X) = \frac{1}{p}$  &  $V(X) = \frac{1-p}{p^2}$ .
4. **Poisson:**  $P(X_\lambda = k) = e^{-\lambda} \cdot \frac{\lambda^k}{k!}$ ,  $E(X_\lambda) = \lambda = V(X_\lambda)$ .
5. **Exponential:**  $P(T \leq t) = F(t) = 1 - e^{-\lambda t}$ ,  $E(T) = \frac{1}{\lambda}$  &  $V(T) = \frac{1}{\lambda^2}$ .
6. **Standard normal distribution:**  $Z = \frac{X - \mu}{\sigma}$ , is a N.D. with  $\mu_Z = 0$  &  $\sigma_Z = 1$ .
7. **Sampling & Central L. T.:** “As the sample size  $n$  increases the shape; of the distribution of sample means from a population “normal or  $n \geq 30$ ” with  $\mu$  and  $\sigma$ ; will approach a N.D. with  $\mu_{\bar{X}} = \mu$  and  $\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$ .” And we use  $z = \frac{\bar{X} - \mu}{\sigma} \cdot \sqrt{n}$ .
8. **Binomial  $\approx$  N.D.:** If a discrete binomial probability distribution satisfies both:  $n \cdot p \geq 5$  and  $n \cdot q \geq 5$ , then it can be approximated by a continuous N.D. with mean  $\mu = n \cdot p$  and a standard deviation  $\sigma = \sqrt{npq}$ .
9. **Confidence interval for  $\mu$ :** If  $\sigma$  is given, then
  - **2-sided:**  $\bar{X} - z_{\frac{\alpha}{2}} \cdot \frac{\sigma}{\sqrt{n}} < \mu < \bar{X} + z_{\frac{\alpha}{2}} \cdot \frac{\sigma}{\sqrt{n}}$
  - **1-sided:**  $\mu > \bar{X} - z_\alpha \cdot \frac{\sigma}{\sqrt{n}} \Leftrightarrow \text{“lower”}$  &  $\mu < \bar{X} + z_\alpha \cdot \frac{\sigma}{\sqrt{n}} \Leftrightarrow \text{“upper”}$
10. **Confidence interval for  $\sigma^2$ :**
  - **2-sided:**  $\frac{(n-1)s^2}{\chi_{\frac{\alpha}{2}}^2} < \sigma^2 < \frac{(n-1)s^2}{\chi_{1-\frac{\alpha}{2}}^2}$
  - **1-sided:**  $\sigma^2 > \frac{(n-1)s^2}{\chi_\alpha^2} \Leftrightarrow \text{“lower”}$  &  $\sigma^2 < \frac{(n-1)s^2}{\chi_{1-\alpha}^2} \Leftrightarrow \text{“upper”}$
11. **Hypothesis Testing: p-value** =  $P(|Z| > |z_0|)$ , where  $z_0$  is the test value
  - (i) For **2-tailed** test:  $p\text{-value} = 2[1 - \Phi(|z_0|)]$
  - (ii) For **left-tailed**:  $p\text{-value} = \Phi(z_0)$ . For **right-tailed**:  $p\text{-value} = 1 - \Phi(z_0)$
12. The **test power** is “ $1 - \beta$ ”, where  $\beta = \Phi(z_{\alpha/2} - \frac{\delta\sqrt{n}}{\sigma}) - \Phi(-z_{\alpha/2} - \frac{\delta\sqrt{n}}{\sigma})$   
 for **2-tailed** test. A sample size that guarantees this **power** is  $n \approx \frac{(z_\beta + z_{\alpha/2})^2 \sigma^2}{\delta^2}$ .  
 For **1-tailed**: if  $\delta > 0$ ,  $\beta = \Phi(z_\alpha - \frac{\delta\sqrt{n}}{\sigma})$ . If  $\delta < 0$ ,  $\beta = 1 - \Phi(-z_\alpha - \frac{\delta\sqrt{n}}{\sigma})$ .