

Lecture Notes for

## ENGI 4421 Probability & Statistics

by Dr. G.H. George Associate Professor, Faculty of Engineering and Applied Science Seventh Edition, reprinted 2021 Spring

http://www.engr.mun.ca/~ggeorge/4421/

## **Table of Contents**

1.	Descriptive Statistics
2.	Introduction to Probability
3.	Counting Techniques
4.	Laws of Probability; Bayes' Theorem
5.	Discrete Random Quantities; Expected Value and Variance
6.	<b>Continuous Random Quantities</b>
7.	Linear Combinations of Random Quantities; Joint Probability Distributions; Correlation; Point Estimation; Bias and Precision
8.	Propagation of Error
9.	Discrete Probability Distributions
10.	<b>Continuous Probability Distributions</b>
11.	<b>Central Limit Theorem; One-sample Confidence Intervals</b>
12.	Two-sample Confidence Intervals
13.	Hypothesis Tests
14.	Chi-Square Tests
15.	Simple Linear Regression
Appendices:	

Appendices:16.Suggestions for Formula Sheets17.Statistical Tables

## List of Symbols:

proper subset  $\subset$  $\subseteq$ subset  $\bigcap$ ,  $\land$ ,  $\times$  intersection (same as logical AND)  $\bigcup$ ,  $\lor$  union = logical inclusive OR empty set = null set = {} Ø probability of type I error α b(x;n,p) binomial p.m.f. B(x;n,p) binomial c.d.f. probability of type II error β  $\beta(\mu_1)$  probability of type II error if  $\mu = \mu_1$ intercept of regression line  $\beta_0$  $\hat{\beta}_0$ estimate of  $\beta_0$  $\beta_1$ slope of regression line  $\hat{\beta}_1$ estimate of  $\beta_1$ CI confidence interval boundary of rejection region or С CI (one-sided) lower boundary of two-sided  $C_L$ rejection region or CI upper boundary of two-sided  $C_U$ rejection region or CI  $^{n}C_{r}$ number of combinations of robjects from nCov[X,Y] covariance  $\chi_{\nu}^{2}$ chi-square distribution with vdegrees of freedom  $\chi^2_{\alpha,v}$  c for which  $P[\chi^2_v > c] = \alpha$ Ε an event (value 1 if true, else 0)  $\tilde{E}$  or  $\sim E$  or not-*E* or  $E^c$  or E' or  $E^*$ or  $\overline{E}$  the complement of event *E* E[X] expected value of X ( =  $\mu$ )

e <sub>i</sub>	residual for <i>i</i> <sup>th</sup> point (in SLR) <b>or</b>
e <sub>i</sub>	number expected in cell $i$ if the
	null hypothesis is true ( $\chi^2$ tests)
e <sub>ii</sub>	number expected in cell $(i, j)$ if
, j	the null hypothesis is true
$\varepsilon^{}_i$	true error for $i^{th}$ point (in SLR)
f	value of <i>F</i> statistic = MSR/MSE
$f_i$	frequency (number of
	observations in $i^{\text{th}}$ interval)
f(x)	probability density function (pdf)
F(x)	cumulative distribution function
	$(\mathrm{cdf}) = \mathrm{P}[X \le x]$
$\Gamma(x)$	gamma function (= ( <i>x</i> –1)!)
$\Gamma(r,\lambda)$ gamma distribution	
$\mathcal{H}_{o}$	null hypothesis
$\mathcal{H}_{A}$ or	$\mathcal{H}_{a}$ or $\mathcal{H}_{1}$ alternative hypothesis
IQR	interquartile range
λ	occurrence rate (in Poisson and
~	exponential distributions)
x ~	sample median
$\mu$	population median
MAD	mean absolute deviation from the
MSE	mean square error $= s^2$
MSR	mean square regression
$\overline{X}$	sample mean (estimator)
$\overline{x}$	sample mean (estimate)
μ	population mean
$\mu^{*}$	posterior estimate of $\mu$
-	(Bayesian)
$\mu_{ m o}$	prior estimate of $\mu$ (Bayesian)
or	value of $\mu$ if null hypothesis true
п	sample size

## List of Symbols (continued)

Ν population size  $N(\mu, \sigma^2)$  normal distribution with mean  $\mu$ , variance  $\sigma^2 v$ number of degrees of freedom number observed in cell i  $O_i$ number observed in cell (i, j) $o_{ij}$ number observed in row *i*  $O_{i\bullet}$ number observed in column *j*  $O_{\bullet i}$ total number observed = n0... population proportion р p = P[E] probability that event E occurs P[A|B] conditional probability (that event A occurs given that event *B* has occurred)  $P[A \cap B]$  joint probability (that both events A and B occur); it is also  $P[A \land B] = P[A \times B] = P[A \text{ and } B] = P[AB]$  $P[A \cup B]$  probability that at least one of events A and B occurs; it is also  $P[A \lor B] = P[A \text{ or } B]$ p(x) = P[X = x] probability mass function (pmf) for discrete x $p(x, y) = P[(X = x) \cap (Y = y)]$  $p_{x}(x) = \sum_{y} p(x, y)$  marginal pmf  $p_{Y|X}(y|x) = p(x,y)/p_X(x)$ Р sample proportion (estimator) p sample proportion (estimate)  $p^*$ adjusted  $\hat{p}$  (Agresti-Coull CI)  $^{n}P_{r}$ number of permutations of robjects from n

ΡI prediction interval (SLR)  $\phi(z)$ standard normal p.d.f.  $\Phi(z)$  standard normal c.d.f.  $q = \tilde{p} = 1 - p$  complement of the probability p Q = 1 - P complement of P  $\hat{q} = 1 - \hat{p}$  complement of  $\hat{p}$  $q_L$  or  $x_L$  lower quartile  $q_U$  or  $x_U$  upper quartile  $q^*$ adjusted  $\hat{q}$  (Agresti-Coull CI) sample correlation or r  $r = \frac{p}{1-p}$  odds that an event with probability p occurs.  $r_i = f_i / n$ relative frequency population correlation ρ S sample space = universal set = possibility space (set of all possible outcomes) sample standard deviation S  $s^2$ sample variance  $\sigma$ population standard deviation  $\sigma^{2}$ population variance  $\sigma^2_{\circ}$ prior variance (Bayesian)  $(\sigma^2)^*$  posterior estimate of  $\sigma^2$ (Bayesian)  $S_{xy} = n \sum xy - \sum x \cdot \sum y$  $S_{xx} = n \sum x^2 - \left(\sum x\right)^2$  $S_{yy} = n \sum y^2 - \left(\sum y\right)^2$ *SIOR* semi-interquartile range simple linear regression SLR standard error  $= \sigma / \sqrt{n}$ s.e. SSE sum of squares due to error sum of squares due to regression SSR SST total sum of squares

random quantity following t - $T_{V}$ distribution with v degrees of freedom t for which  $P[T_v > t] = \alpha$  $t_{\alpha,\nu}$ observed value of t $t_{\rm obs}$ V[X] variance of  $X = \sigma^2$ prior weight (Bayesian) w<sub>o</sub> data weight (Bayesian)  $W_d$ value of *y* predicted from ŷ regression line Ζ standard normal random quantity z for which  $P[Z > z] = \alpha$  $z_{\alpha}$ observed value of z $Z_{\rm obs}$