

## Engineering Statistics

### STATISTICAL FORMULAE

1. For  $X$  having the Binomial distribution,  $B(n, p)$ :  $P(X = k) = \binom{n}{k} p^k (1-p)^{n-k}$ ,  
for  $k = 0, 1, 2, \dots, n$  and  $X$  has mean  $np$  and variance  $np(1-p)$ .
2. For  $Y$  having the Poisson distribution, parameter  $\lambda$ :  $P(Y = k) = \frac{e^{-\lambda} \lambda^k}{k!}$ ,  
for  $k = 0, 1, 2, \dots$  and  $Y$  has mean  $\lambda$  and variance  $\lambda$ .
3. For the sample  $x_1, x_2, \dots, x_n$ , the sample mean is  $\bar{x} = \sum_{i=1}^n x_i / n$   
and the sample variance is  $s^2 = \sum_{i=1}^n (x_i - \bar{x})^2 / (n-1) = \left( \sum_{i=1}^n x_i^2 - n\bar{x}^2 \right) / (n-1)$ .
4. Observations  $x_1, x_2, \dots, x_m$  occur with frequencies  $f_1, f_2, \dots, f_m$ , the sample mean  
is  $\bar{x} = \frac{\sum_{i=1}^m f_i x_i}{n}$  and the sample variance is  $s^2 = \frac{\sum_{i=1}^m f_i (x_i - \bar{x})^2}{n-1} = \frac{\sum_{i=1}^m f_i x_i^2 - n\bar{x}^2}{n-1}$ .
5. Random sample  $X_1, X_2, \dots, X_n$  from a  $N(\mu, \sigma^2)$  distribution, then  $\frac{\bar{X} - \mu}{\sqrt{\sigma^2/n}}$  has a  
 $N(0, 1)$  distribution and  $\frac{\bar{X} - \mu}{\sqrt{s^2/n}}$  has a  $t_{n-1}$  distribution.
6.  $B(n, p)$  is approximated by  $N(np, np(1-p))$ , when  $n$  is large and  $np$  is not too close  
to 0 or  $n$ . Poisson,  $\lambda$ , is approximated by  $N(\lambda, \lambda)$ , when  $\lambda$  is large.
7. The linear regression line is estimated by  $y = \hat{a} + \hat{b}x$  where  $\hat{b} = S_{xy} / S_{xx}$ ,  
 $\hat{a} = \bar{y} - \hat{b}\bar{x}$ ,  $\hat{\sigma}^2 = \frac{S_{yy} - \hat{b}S_{xy}}{n-2}$ ,  $S_{xx} = \sum x_i^2 - \frac{(\sum x_i)^2}{n}$ ,  $S_{yy} = \sum y_i^2 - \frac{(\sum y_i)^2}{n}$  and  
 $S_{xy} = \sum x_i y_i - \frac{\sum x_i \sum y_i}{n}$ .  $\frac{\hat{b} - b}{\sqrt{\hat{\sigma}^2 / S_{xx}}}$  has a  $t_{n-2}$  distribution. The mean value of  $y$   
at  $x_0$ ,  $a + bx_0$ , has confidence interval  $\hat{a} + \hat{b}x_0 \pm t_{n-2} \sqrt{\hat{\sigma}^2 \left( \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right)}$ .  
A confidence interval for a single response at  $x_0$  is  
 $\hat{a} + \hat{b}x_0 \pm t_{n-2} \sqrt{\hat{\sigma}^2 \left( 1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}} \right)}$ .  
The (Pearson product-moment) sample correlation is  $r = \frac{S_{xy}}{\sqrt{S_{xx} S_{yy}}}$ .