

# BSTA 320 – COMPREHENSIVE EXAM FORMULA SHEET

## ***Decision Under Uncertainty***

Maximax

Maximin

Equally Likely (Laplace)

Criterion of Realism (Hurwicz):

$$\alpha \times (\text{best payoff for an alternative}) + (1 - \alpha) \times (\text{worst payoff for the alternative})$$

Minimax Regret

## ***Decision Making Under Risk***

(Outcomes are also known as States of Nature)

### **Expected Monetary Value**

$$\begin{aligned} \text{EMV} = & (\text{payoff of first outcome}) \times (\text{probability of first outcome}) \\ & + (\text{payoff of second outcome}) \times (\text{probability of second outcome}) \\ & + \dots + (\text{payoff of last outcome}) \times (\text{probability of last outcome}) \end{aligned}$$

### **Expected Opportunity Loss**

$$\begin{aligned} \text{EOL} = & (\text{regret of first outcome}) \times (\text{probability of first outcome}) \\ & + (\text{regret of second outcome}) \times (\text{probability of second outcome}) \\ & + \dots + (\text{regret of last outcome}) \times (\text{probability of last outcome}) \end{aligned}$$

### **Expected Value with Perfect Information**

$$\begin{aligned} \text{EVwPI} = & (\text{best payoff of the first outcome}) \times (\text{probability of first outcome}) \\ & + (\text{best payoff of the second outcome}) \times (\text{probability of second outcome}) \\ & + \dots + (\text{best payoff of the last outcome}) \times (\text{probability of last outcome}) \end{aligned}$$

### **Expected Value of Perfect Information**

$$\text{EVPI} = \text{EVwPI} - \text{Max EMV}$$

Max EMV: The expected value without information.

## ***Decision Making with Sample Information***

### **Expected Value of Sample Information**

$$\text{EVSI} = \left( \begin{array}{l} \text{expected value of best decision} \\ \text{with sample information,} \\ \text{assuming no cost to gather it} \end{array} \right) - \left( \begin{array}{l} \text{expected value of} \\ \text{best decision without} \\ \text{sample information} \end{array} \right)$$

$$= \text{EVwSI} - \text{Max EMV}$$

$$\text{Efficiency} = \frac{\text{EVSI}}{\text{EVPI}}$$

**Table: Computation of Posterior Probabilities**

(1) Outcome	(2) Prior Probabilities	(3) Conditional Probabilities	(4) Joint Probabilities (2) x (3)	(5) Posterior Probabilities (4) / $\sum(4)$

**BAYES' THEOREM (for calculation of Posterior Probabilities)**

The probability of event  $B_i$  given that event  $A$  has occurred is given by the formula

$$P(B_i|A) = \frac{P(B_i)P(A|B_i)}{P(B_1)P(A|B_1)+P(B_2)P(A|B_2) + \dots +P(B_k)P(A|B_k)}$$

where  $B_1, B_2, \dots, B_k$  are **mutually exclusive** and **collectively exhaustive** events.

**Control Charts**

<b>R chart</b>	<b><math>\bar{x}</math> chart</b>
$UCL = \bar{R} D_4$ $LCL = \bar{R} D_3$ <p>where <math>\bar{R} = \frac{\sum R}{k}</math>                      k= number of sub groups sampled</p>	$UCL = \bar{\bar{x}} + A_2 \bar{R}$ $LCL = \bar{\bar{x}} - A_2 \bar{R}$ <p>where <math>\bar{\bar{x}} = \frac{\sum \bar{x}}{k}</math></p>
<b>p chart</b>	<b>c chart</b>
<p><b>Control Limits:</b> <math>\bar{p} \pm 3 \sqrt{\frac{(\bar{p})(\bar{q})}{\bar{n}}}</math></p> <p>where <math>\bar{p} = \frac{\sum x}{\sum n}</math>, <math>\bar{q} = 1 - \bar{p}</math>, <math>\bar{n} = \frac{\sum n}{k}</math>                      where <math>k</math> = number of subgroups sampled</p>	<p><b>Control Limits:</b> <math>\bar{c} \pm 3 \sqrt{\bar{c}}</math></p> <p>where <math>\bar{c} = \frac{\sum c}{k}</math>  <math>c</math> = number of occurrences and  <math>k</math> = number of units sampled</p>

## TABLE OF CONTROL CHART CONSTANTS

<i>OBSERVATIONS IN SAMPLE, n</i>	<i>CHART FOR AVERAGES</i>	<i>CHART FOR RANGES</i>	<i>FACTORS FOR CONTROL LIMITS</i>	
	$\bar{x}$	$\bar{R}$	$A_2$	$A_3$
	$A_2$	$A_3$	$D_3$	$D_4$
2	1.880	2.659	0	3.267
3	1.023	1.954	0	2.575
4	0.729	1.628	0	2.282
5	0.577	1.427	0	2.114
6	0.483	1.287	0	2.004
7	0.419	1.182	0.076	1.924
8	0.373	1.099	0.136	1.864
9	0.337	1.032	0.184	1.816
10	0.308	0.975	0.223	1.777
11	0.285	0.927	0.256	1.744
12	0.266	0.886	0.283	1.717
13	0.249	0.850	0.307	1.693
14	0.235	0.817	0.328	1.672
15	0.223	0.789	0.347	1.653
16	0.212	0.763	0.363	1.637
17	0.203	0.739	0.378	1.622
18	0.194	0.718	0.391	1.609
19	0.187	0.698	0.404	1.596
20	0.180	0.680	0.415	1.585
21	0.173	0.663	0.425	1.575
22	0.167	0.647	0.435	1.565
23	0.162	0.633	0.443	1.557
24	0.157	0.619	0.452	1.548
25	0.153	0.606	0.459	1.541

Source: Adapted from *Manual on Presentation of Data and Control Chart Analysis*, Copyright ©2002 by ASTM International, p. 67

## Forecasting

### *Moving Average*

$$k\text{-period moving average} = \sum (\text{actual value in previous } k \text{ periods})/k$$

### *Weighted Moving Average*

$$k\text{-period weighted moving average} = \frac{\sum_i^k (\text{weight for period } i) \times (\text{Actual value in period } i)}{\sum_i^k (\text{weights})}$$

## Exponential Smoothing

$$F_{t+1} = F_t + \alpha (A_t - F_t) \quad \text{or} \quad F_{t+1} = \alpha A_t + (1 - \alpha)F_t$$

where  $A_t$  = actual value in period  $t$ ,  $F_t$  = forecast for period  $t$

$F_{t+1}$  = forecast for period  $t + 1$ ,

$\alpha$  = smoothing constant value ( $0 \leq \alpha \leq 1$ )

$$MAD = \sum_{t=1}^T | \text{forecast error} | / T$$

$$MAPE = 100 \sum_{t=1}^T (| \text{forecast error} | / A_t) / T$$

## Simple Linear Regression

$$\hat{y} = a + b x \quad \text{or} \quad \hat{y} = b_0 + b_1 x$$

**Coefficient of Correlation ( $r$ )** measures the strength of the relationship ( $-1 \leq r \leq 1$ )

$$\text{Coefficient of Determination } (R^2) = \frac{\text{explained variation}}{\text{total variation}} \text{ or } \frac{SSR}{SST}$$

where  $SST = SSR + SSE$

$SST$  = total sum of squares,  $SSR$  = regression sum of squares,  $SSE$  = error sum of squares

$$\text{Standard Error of Estimate } (s_e) \quad s_e = \sqrt{\frac{SSE}{df}} \quad \text{or} \quad s_e = \sqrt{\frac{SSE}{n-2}}$$

## Multiple Regression

$$\hat{y} = b_0 + b_1 x_1 + b_2 x_2 + \dots + b_k x_k$$

$$\text{Standard Error of Estimate } (s_e) \quad s_e = \sqrt{\frac{SSE}{df}} \quad \text{or} \quad s_e = \sqrt{\frac{SSE}{n - (k + 1)}} \quad \text{or} \quad s_e = \sqrt{\frac{\sum (y - \hat{y})^2}{n - (k + 1)}}$$

where  $\hat{y}$  = predicted  $y$  value and  $y$  = actual  $y$  value

$df = n - (k + 1)$

$df$  = degrees of freedom,  $k$  = number of independent ( $x$ ) variables,  $n$  = sample size

$$\text{Coefficient of Determination } (R^2) = \frac{\text{explained variation}}{\text{total variation}} \text{ or } \frac{SSR}{SST}$$

$$\text{Adjusted } R^2 = 1 - \frac{(n-1)}{[n - (k + 1)]} (1 - R^2)$$

where  $n$  = sample size and  $k$  = number of independent ( $x$ ) variables