

$M_1 - M_2 = (M_1 + M_3) - (M_2 + M_3) = M_1 - M_2$   
 $M_1 + M_2 = M_1 + M_2 = M_3 + M_4$

Low	Med	High
$M_1$	$M_2$	$M_3$
$M_4$	$M_5$	$M_6$

$L = \begin{pmatrix} 1 & -1 & 0 & 1 & -1 & 0 \\ 0 & 1 & -1 & 0 & 1 & -1 \end{pmatrix}$

	L	L	1	0	1	0
2	L	M	-1	1	-1	1
3	L	H	0	-1	0	-1
4	H	L	1	0	-1	0
5	H	M	-1	1	1	-1
6	H	H	0	-1	0	1

$Y = \beta_0 + \beta_1 d_1 + \beta_2 d_2 + \beta_3 d_3 + \beta_4 d_4 + \beta_5 d_5 + \beta_6 d_6 + \epsilon$

Main effect of A  
 $H_0: \beta_1 = 0$   
 $H_0: \beta_2 = 0$   
 $H_0: \beta_3 = 0$

A	B	d	d <sub>1</sub>	E(Y)
1	1	1	1	
1	2	1	-1	
2	1	-1	1	
2	2	-1	-1	

B  

A	B	d	d <sub>1</sub>	E(Y)
1	1	1	1	$\beta_0 + \beta_1 + \beta_2 + \beta_3$
1	2	1	0	$\beta_0 + \beta_1$
2	1	0	1	$\beta_0 + \beta_2$
2	2	0	0	$\beta_0$

	1	2
1	$\beta_0 + \beta_1 + \beta_2$	$\beta_0 + \beta_1$
2	$\beta_0 + \beta_2$	$\beta_0$

$\beta_0 + \beta_1 + \beta_2$      $\beta_0 + \beta_1$   
 $\beta_0 + \beta_2$      $\beta_0$

5 a)  $Y_{ij} \sim N(\mu, \sigma^2)$   
 b) Within school: random student selection  
 Outside school: random school selection  $\rightarrow Y_{ij}$  indep.  
 $Cov(Y_{ij}, Y_{j'}) = Cov(M + Y_i + \epsilon_{ij}, M + Y_i + \epsilon_{j'}) = E(\epsilon_{ij} \epsilon_{j'}) = 0$   
 $Cov(Y_{ij}, Y_{j'}) = E(\epsilon_{ij} \epsilon_{j'}) = 0$   
 $N(\mu, \frac{\sigma^2}{k})$   
 $Var(\bar{Y}_i) = Var(\frac{1}{k} \sum_{j=1}^k Y_{ij}) = \frac{1}{k^2} Var(\sum_{j=1}^k (\mu + Y_i + \epsilon_{ij})) = \frac{1}{k^2} Var(\sum_{j=1}^k \epsilon_{ij}) = \frac{1}{k^2} (k \sigma^2) = \frac{\sigma^2}{k}$   
 $\bar{Y}_i = \mu + Y_i + \bar{\epsilon}_i$   
 $Cov(\bar{Y}_i, \bar{Y}_{j'}) = Cov(\mu + Y_i + \bar{\epsilon}_i, \mu + Y_i + \bar{\epsilon}_{j'}) = Cov(\bar{\epsilon}_i, \bar{\epsilon}_{j'}) = 0$   
 $E\{(\mu + Y_i + \bar{\epsilon}_i)(\mu + Y_i + \bar{\epsilon}_{j'})\} = Cov(\bar{\epsilon}_i, \bar{\epsilon}_{j'}) = 0$

6  $\frac{\sum_{i=1}^g (\bar{Y}_i - \bar{Y})^2}{\frac{\sigma^2 + k\sigma^2}{k}} \sim \chi^2(g-1)$   
 $E(\frac{SST}{\sigma^2 + k\sigma^2}) = \frac{g-1}{k} \cdot k \sigma^2 = (g-1) \sigma^2$   
 $SST = \sum_{i=1}^g \sum_{j=1}^k (Y_{ij} - \bar{Y})^2 = \sum_{i=1}^g \sum_{j=1}^k (Y_{ij} - \bar{Y}_i + \bar{Y}_i - \bar{Y})^2 = \sum_{i=1}^g \sum_{j=1}^k (Y_{ij} - \bar{Y}_i)^2 + \sum_{i=1}^g \sum_{j=1}^k (\bar{Y}_i - \bar{Y})^2 = \sum_{i=1}^g \sum_{j=1}^k (Y_{ij} - \bar{Y}_i)^2 + (g-1) \sigma^2$   
 $\frac{\sum_{i=1}^g \sum_{j=1}^k (Y_{ij} - \bar{Y}_i)^2}{\sigma^2 + k\sigma^2} \sim \chi^2(g-1)$   
 7  $SSE = \sum_{i=1}^g \sum_{j=1}^k \frac{(Y_{ij} - \bar{Y}_i)^2}{\sigma^2} \sim \chi^2(g(k-1))$   
 8 ESTIMATE  $\frac{\sigma^2}{\sigma^2 + \sigma^2}$  Expected MS  
 $E(MSE) = \sigma^2$   
 $E(MST) = \sigma^2 + k\sigma^2$  if  $\sigma^2 = 0$   
 9  $F = \frac{SST / (g-1)}{SSE / (g(k-1))} \sim F(g-1, g(k-1))$   
 $\sum_{i=1}^g (\bar{Y}_i - \bar{Y})^2 \sim g_1 (\bar{Y}_1 - \bar{Y}_g)$   
 $\sum_{i=1}^g \sum_{j=1}^k (Y_{ij} - \bar{Y}_i)^2 \sim g_2 (S_{11} - S_g)$   
 $\bar{Y}_i$  ind of  $S_i$   
 $\bar{Y}_i$  ind of  $S_i, i \neq j$