

For Each Problem, assume the Completely Randomized Design (1-Way ANOVA):

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$$Y_{ij} = \mu_i + e_{ij} = \mu + \tau_i + e_{ij} \quad i=1, \dots, t; j=1, \dots, r_i \quad e_{ij} \sim NID(0, \sigma^2) \quad \sum_{i=1}^t r_i \tau_i = 0$$

F[A=.95]	df1								Studentized Range								Bonferroni t				
df2	1	2	3	4	5	6	7	8	df#trts	2	3	4	5	6	7	8	df#Comparis	1	3	6	10
1	161.448	199.500	215.707	224.583	230.162	233.986	236.768	238.883	1	17.969	26.976	32.819	37.082	40.408	43.119	45.397	1	12.706	38.188	76.390	127.321
2	18.513	19.000	19.164	19.247	19.296	19.330	19.353	19.371	2	6.085	8.331	9.798	10.881	11.734	12.435	13.027	2	4.303	7.649	10.886	14.089
3	10.128	9.552	9.277	9.117	9.013	8.941	8.887	8.845	3	4.501	5.910	6.825	7.502	8.037	8.478	8.852	3	3.182	4.857	6.232	7.453
4	7.709	6.944	6.591	6.388	6.256	6.163	6.094	6.041	4	3.926	5.040	5.757	6.287	6.706	7.053	7.347	4	2.776	3.961	4.851	5.598
5	6.608	5.786	5.409	5.192	5.050	4.950	4.876	4.818	5	3.635	4.602	5.218	5.673	6.033	6.330	6.582	5	2.571	3.534	4.219	4.773
6	5.987	5.143	4.757	4.534	4.387	4.284	4.207	4.147	6	3.460	4.339	4.896	5.305	5.628	5.895	6.122	6	2.447	3.287	3.863	4.317
7	5.591	4.737	4.347	4.120	3.972	3.866	3.787	3.726	7	3.344	4.165	4.681	5.060	5.359	5.606	5.815	7	2.365	3.128	3.636	4.029
8	5.318	4.459	4.066	3.838	3.687	3.581	3.500	3.438	8	3.261	4.041	4.529	4.886	5.167	5.399	5.596	8	2.306	3.016	3.479	3.833
9	5.117	4.256	3.863	3.633	3.482	3.374	3.293	3.230	9	3.199	3.948	4.415	4.755	5.024	5.244	5.432	9	2.262	2.933	3.364	3.690
10	4.965	4.103	3.708	3.478	3.326	3.217	3.135	3.072	10	3.151	3.877	4.327	4.654	4.912	5.124	5.304	10	2.228	2.870	3.277	3.581
11	4.844	3.982	3.587	3.357	3.204	3.095	3.012	2.948	11	3.113	3.820	4.256	4.574	4.823	5.028	5.202	11	2.201	2.820	3.208	3.497
12	4.747	3.885	3.490	3.259	3.106	2.996	2.913	2.849	12	3.081	3.773	4.199	4.508	4.750	4.950	5.119	12	2.179	2.779	3.153	3.428
13	4.667	3.806	3.411	3.179	3.025	2.915	2.832	2.767	13	3.055	3.734	4.151	4.453	4.690	4.884	5.049	13	2.160	2.746	3.107	3.372
14	4.600	3.739	3.344	3.112	2.958	2.848	2.764	2.699	14	3.033	3.701	4.111	4.407	4.639	4.829	4.990	14	2.145	2.718	3.069	3.326
15	4.543	3.682	3.287	3.056	2.901	2.790	2.707	2.641	15	3.014	3.673	4.076	4.367	4.595	4.782	4.940	15	2.131	2.694	3.036	3.286
16	4.494	3.634	3.239	3.007	2.852	2.741	2.657	2.591	16	2.998	3.649	4.046	4.333	4.557	4.741	4.896	16	2.120	2.673	3.008	3.252
17	4.451	3.592	3.197	2.965	2.810	2.699	2.614	2.548	17	2.984	3.628	4.020	4.303	4.524	4.705	4.858	17	2.110	2.655	2.984	3.222
18	4.414	3.555	3.160	2.928	2.773	2.661	2.577	2.510	18	2.971	3.609	3.997	4.276	4.494	4.673	4.824	18	2.101	2.639	2.963	3.197
19	4.381	3.522	3.127	2.895	2.740	2.628	2.544	2.477	19	2.960	3.593	3.977	4.253	4.468	4.645	4.794	19	2.093	2.625	2.944	3.174
20	4.351	3.493	3.098	2.866	2.711	2.599	2.514	2.447	20	2.950	3.578	3.958	4.232	4.445	4.620	4.768	20	2.086	2.613	2.927	3.153
21	4.325	3.467	3.072	2.840	2.685	2.573	2.488	2.420	21	2.941	3.565	3.942	4.213	4.424	4.597	4.743	21	2.080	2.601	2.912	3.135
22	4.301	3.443	3.049	2.817	2.661	2.549	2.464	2.397	22	2.933	3.553	3.927	4.196	4.405	4.577	4.722	22	2.074	2.591	2.899	3.119
23	4.279	3.422	3.028	2.796	2.640	2.528	2.442	2.375	23	2.926	3.542	3.914	4.180	4.388	4.558	4.702	23	2.069	2.582	2.886	3.104
24	4.260	3.403	3.009	2.776	2.621	2.508	2.423	2.355	24	2.919	3.532	3.901	4.166	4.373	4.541	4.684	24	2.064	2.574	2.875	3.091
25	4.242	3.385	2.991	2.759	2.603	2.490	2.405	2.337	25	2.913	3.523	3.890	4.153	4.358	4.526	4.667	25	2.060	2.566	2.865	3.078
26	4.225	3.369	2.975	2.743	2.587	2.474	2.388	2.321	26	2.907	3.514	3.880	4.141	4.345	4.511	4.652	26	2.056	2.559	2.856	3.067
27	4.210	3.354	2.960	2.728	2.572	2.459	2.373	2.305	27	2.902	3.506	3.870	4.130	4.333	4.498	4.638	27	2.052	2.552	2.847	3.057
28	4.196	3.340	2.947	2.714	2.558	2.445	2.359	2.291	28	2.897	3.499	3.861	4.120	4.322	4.486	4.625	28	2.048	2.546	2.839	3.047
29	4.183	3.328	2.934	2.701	2.545	2.432	2.346	2.278	29	2.892	3.493	3.853	4.111	4.311	4.475	4.613	29	2.045	2.541	2.832	3.038
30	4.171	3.316	2.922	2.690	2.534	2.421	2.334	2.266	30	2.888	3.486	3.845	4.102	4.301	4.464	4.601	30	2.042	2.536	2.825	3.030
40	4.085	3.232	2.839	2.606	2.449	2.336	2.249	2.180	40	2.858	3.442	3.791	4.039	4.232	4.388	4.521	40	2.021	2.499	2.776	2.971
50	4.034	3.183	2.790	2.557	2.400	2.286	2.199	2.130	48	2.843	3.420	3.764	4.008	4.197	4.351	4.481	50	2.009	2.477	2.747	2.937
60	4.001	3.150	2.758	2.525	2.368	2.254	2.167	2.097	60	2.829	3.399	3.737	3.977	4.163	4.314	4.441	60	2.000	2.463	2.729	2.915
70	3.978	3.128	2.736	2.503	2.346	2.231	2.143	2.074	80	2.814	3.377	3.711	3.947	4.129	4.277	4.402	70	1.994	2.453	2.715	2.899
80	3.960	3.111	2.719	2.486	2.329	2.214	2.126	2.056									80	1.990	2.445	2.705	2.887

Q.1 Derive $E\{MS_{Treats}\}$ for the balanced 1-Way ANOVA with t treatments and r replicates per treatment by completing the following parts. **SHOW ALL WORK to Receive any Credit!!!!!!!**

p.1.a. Show: $SS_{Treats} = \sum_{i=1}^t \sum_{j=1}^r (\bar{y}_{i.} - \bar{y}_{..})^2 = r \sum_{i=1}^t (\bar{y}_{i.})^2 - rt(\bar{y}_{..})^2$ 12

$$= \sum_i \sum_j (\bar{y}_{i.}^2 + \bar{y}_{..}^2 - 2\bar{y}_{i.}\bar{y}_{..}) = \sum_i \sum_j \bar{y}_{i.}^2 + \sum_i \sum_j \bar{y}_{..}^2 - 2 \sum_i \sum_j \bar{y}_{i.}\bar{y}_{..}$$

$$= r \sum_i \bar{y}_{i.}^2 + tr \bar{y}_{..}^2 - 2r \bar{y}_{..} \sum_i \bar{y}_{i.}$$

$$\frac{\sum_i \bar{y}_{i.}}{t} = \bar{y}_{..} \Rightarrow \sum_i \bar{y}_{i.} = t \bar{y}_{..}$$

$$= r \sum_i \bar{y}_{i.}^2 + tr \bar{y}_{..}^2 - 2tr \bar{y}_{..}^2$$

$$= r \sum_i \bar{y}_{i.}^2 - tr \bar{y}_{..}^2$$

p.1.b. Derive $E\{\bar{y}_{i.}\}$, $V\{\bar{y}_{i.}\}$, $E\{(\bar{y}_{i.})^2\}$, $E\{\bar{y}_{..}\}$, $V\{\bar{y}_{..}\}$, $E\{(\bar{y}_{..})^2\}$

$$E\{y_{ij}\} = \mu_i \quad V\{y_{ij}\} = \sigma^2 \quad \text{Cov}\{y_{ij}, y_{i'j'}\} = 0 \quad \forall i \neq i' \text{ and/or } j \neq j'$$

$$E\left\{\frac{1}{r} \sum_j y_{ij}\right\} = \frac{1}{r} \sum_j E\{y_{ij}\} = \frac{1}{r} r \mu_i = \mu_i \quad V\left\{\frac{1}{r} \sum_j y_{ij}\right\} = \frac{1}{r^2} r \sigma^2 = \frac{\sigma^2}{r}$$

$$\Rightarrow E\{(\bar{y}_{i.})^2\} = \mu_i^2 + \frac{\sigma^2}{r} = (\mu_i^2 + \sigma_i^2 + 2\mu_i \sigma_i) + \frac{\sigma^2}{r}$$

$$E\left\{\frac{1}{tr} \sum_i \sum_j y_{ij}\right\} = \frac{1}{tr} \sum_i \sum_j E\{y_{ij}\} = \frac{1}{tr} \sum_i r \mu_i = \frac{1}{t} \sum_i \mu_i = \bar{\mu}$$

$$V\left\{\frac{1}{tr} \sum_i \sum_j y_{ij}\right\} = \left(\frac{1}{tr}\right)^2 tr \sigma^2 = \frac{\sigma^2}{tr}$$

$$\Rightarrow E\{\bar{y}_{..}^2\} = \bar{\mu}^2 + \frac{\sigma^2}{tr}$$

p.1.c. Use results from p.1.a. and p.1.b. to obtain: $E\{SS_{\text{TRTS}}\}$ and $E\{MS_{\text{TRTS}}\}$

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$$E\{SS_{\text{TRTS}}\} = r \sum_i \left[(\bar{\mu}^2 + \tau_i^2 + 2\bar{\mu} \cdot \tau_i) + \frac{\sigma^2}{r} \right] - t r \left[\bar{\mu}^2 + \frac{\sigma^2}{t r} \right]$$

$$= (t r \bar{\mu}^2 + r \sum_i \tau_i^2 + 2\bar{\mu} \cdot r \sum_i \tau_i + t \sigma^2) - (t r \bar{\mu}^2 + \sigma^2)$$

$$= r \sum_i \tau_i^2 + (t-1) \sigma^2$$

$$\Rightarrow E\{MS_{\text{TRTS}}\} = E\left\{\frac{SS_{\text{TRTS}}}{t-1}\right\} = \frac{r \sum_i \tau_i^2}{t-1} + \sigma^2$$

Q.1.d. For which scenario of the parameter values, will the power of the F-test be highest (assume $t=3$, $r_1=r_2=r_3=r$):

Scenario 1: $\mu_1=80, \mu_2=100, \mu_3=120, \sigma=10$ or **Scenario 2:** $\mu_1=90, \mu_2=100, \mu_3=110, \sigma=5$

Compute $E\{MS_{\text{TRTS}}\}$ for each scenario (as a function of r).

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$$\textcircled{1} \quad \tau_1 = -20 \quad \tau_2 = 0 \quad \tau_3 = 20 \Rightarrow \sum_i \tau_i^2 = 800$$

$$\Rightarrow E\{MS_{\text{TRTS}}\} = \frac{800r}{2} + 100 = 100 + 400r$$

$$\textcircled{2} \quad \tau_1 = -10 \quad \tau_2 = 0 \quad \tau_3 = 10 \Rightarrow \sum_i \tau_i^2 = 200$$

$$E\{MS_{\text{TRTS}}\} = \frac{200r}{2} + 25 = 100r + 25$$

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Q.2. A published report, based on a **balanced 1-Way ANOVA** reports means (SDs) for the three treatments as:

Trt 1: 70 (8) Trt 2: 75 (6) Trt 3: 80 (10) Tragically, the authors fail to give the treatment sample sizes.

p.2.a The Treatment degrees of freedom is the same for each sample size. The error degrees of freedom are:

$$\sum_i (\bar{y}_i - \bar{y}_{..})^2 = 25 + 0 + 25 = 50$$

$$\sum_i s_i^2 = 64 + 36 + 100 = 200$$

$$df_{TRTS} = \overset{3}{3} - 1 = 2 \quad r_i = 2: df_{ERR} = \overset{3}{3}(2-1) = 3 \quad r_i = 6: df_{ERR} = \overset{3}{3}(6-1) = 15 \quad r_i = 10: df_{ERR} = \overset{3}{3}(10-1) = 27$$

p.2.b. Complete the following table, given arbitrary levels of the number of replicates per treatment:

r	SSTrt	SSErr	MSTrt	MSErr	F_obs	F(0.05)
2	2(50) = 100	1(200) = 200	50	66.67	0.75	9.552
6	6(50) = 300	5(200) = 1000	150	66.67	2.25	3.682
10	10(50) = 500	9(200) = 1800	250	66.67	3.75	3.354

p.2.c. The smallest r_i , so that these means are significantly different is:

i) $r_i \leq 2$

ii) $2 < r_i \leq 6$

iii) $6 < r_i \leq 10$

iv) $r_i > 10$

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Q.3. A study is conducted to compare 4 menus in terms of numbers of calories ordered by restaurant customers (in 100s of calories). The treatments (menus) are (consider them increasing in order of information provided):

- 1) No Calories Reported 2) Calories Reported 3) Rank-Ordered Calories 4) Color-Coded Calories

The sample sizes are all based on samples of $r = 20$ customers per menu. The sample means and estimated variance are:

$\bar{Y}_1 = 17.6$ $\bar{Y}_2 = 16.8$ $\bar{Y}_3 = 16.0$ $\bar{Y}_4 = 14.4$ $\bar{Y}_{..} = 16.2$ $s^2 = \text{MSE}_{\text{err}} = 196.0$

p.3.a. Complete the following table to test $H_0: \mu_1 = \mu_2 = \mu_3 = \mu_4$

ANOVA	df	SS	MS	F_obs	F(.05)	Reject H0?
Source						Yes / No
Trts (Menus)	3	48.96 112	32.20 37.33	#N/A	#N/A	#N/A
Error	76	1132096	14876 1490	#N/A	#N/A	#N/A
Total	79		#N/A	#N/A	#N/A	#N/A

$F^* = 0.167$
 $F^* = 0.190$
 $F(.05, 3, 76) \approx 2.72$

$SS_{\text{Trts}} = 20 \left[1.4^2 + 0.6^2 + (-0.2)^2 + (-1.8)^2 \right]$ $SSE = 76(196) = 14896$
 $= 20 \left[1.96 + 0.36 + 0.04 + 3.24 \right] = 20(5.6) = 112$

p.3.b.: Give 3 orthogonal contrasts, and their estimates and estimated standard errors

Contrast 1: Menu 1 vs Menus {2, 3, 4} Contrast 2: Menu 2 vs Menus {3, 4} Contrast 3: Menu 3 vs Menu 4

Contrast 1: $C_1 = 3\mu_1 - \mu_2 - \mu_3 - \mu_4$ $c_1 = 52.8 - 16.8 - 16.0 - 14.4 = 5.6$ $SE[c_1] = \sqrt{196 \left(\frac{9+1+1+1}{20} \right)} = \sqrt{117.6} = 10.84$

Contrast 2: $C_2 = 2\mu_2 - \mu_3 - \mu_4$ $c_2 = 33.6 - 16 - 14.4 = 3.2$ $SE[c_2] = \sqrt{196 \left(\frac{4+1+1}{20} \right)} = \sqrt{58.8} = 7.67$

Contrast 3: $C_3 = \mu_3 - \mu_4$ $c_3 = 16.0 - 14.4 = 1.6$ $SE[c_3] = \sqrt{196 \left(\frac{1+1}{20} \right)} = \sqrt{19.6} = 4.43$

p.3.c. Obtain Simultaneous 95% CIs for each (population) contrast, based on Bonferroni's method

$t_{(\frac{.025}{3}, 76)} \approx 2.448$

Contrast 1: $5.6 \pm 2.448(10.84) \approx 5.6 \pm 26.5 \approx (-20.9, 32.1)$

Contrast 2: $3.2 \pm 2.448(7.67) \approx 3.2 \pm 18.8 \approx (-15.6, 22.0)$

Contrast 3: $1.6 \pm 2.448(4.43) \approx 1.6 \pm 10.8 \approx (-9.2, 12.4)$

Q.4. Players in the English Premier Soccer League are classified as one of 4 positions (Defender, Forward, Goalie, or Midfielder). Random samples of 8 players were selected from each position for players during the 2014 season. The players' Body Mass Indices were measured. The within position (Error) sum of squares is $SSE = 68$. The means for the 4 positions are: 22.77 (D), 22.19 (F), 24.11 (G), and 23.16 (M).

$$df_e = 32 - 4 = 28$$

p.4.a. Compute the standard error of the difference between 2 means: $s\{\bar{Y}_{i.} - \bar{Y}_{j.}\}$ $MSE = \frac{68}{28} = 2.43$

$$\sqrt{2.43\left(\frac{1}{8} + \frac{1}{8}\right)} = \sqrt{.6071} = 0.78$$

$$s\{\bar{Y}_{i.} - \bar{Y}_{j.}\} = \underline{0.78}$$

$$\text{Max diff} = 24.11 - 22.19 = 1.92$$

p.4.b. Compute Tukey's Honest Significant Difference for simultaneously comparing all pairs of positions, with a family-wise error rate of 0.05. Identify significant differences (if any) among all pairs of means.

$$q(.05, 4, 28) = 3.861$$

$$HSD = \frac{3.861}{\sqrt{2}} (0.78) = 2.13$$

$$\text{Tukey's HSD} = \underline{2.13}$$

None significantly different

p.4.c. Compute Bonferroni's Minimum Significant Difference for simultaneously comparing all pairs of positions, with a family-wise error rate of 0.05. Identify significant differences (if any) among all pairs of means.

$$t\left(\frac{.025}{6}, 28\right) = 2.839 \quad B = 2.839(0.78) = 2.21$$

$$\text{Bonferroni's MSD} = \underline{2.21}$$

None different

p.4.d. Compute Scheffe's Minimum Significant Difference for simultaneously comparing all pairs of positions, with a family-wise error rate of 0.05. Identify significant differences (if any) among all pairs of means.

$$\sqrt{(4-1)F(.05, 3, 28)} = \sqrt{3(2.947)} = 2.97$$

$$2.97(0.78) = 2.32$$

$$\text{Scheffe's MSD} = \underline{2.32}$$

None different

Q.5. Consider the following 2 models:

Model 1: $y_{ij} = \mu_i + e_{ij}$ $i=1, \dots, t; j=1, \dots, r_i$ Model 2: $y_{ij} = \mu + e_{ij}$ $i=1, \dots, t; j=1, \dots, r_i$

p.5.a. Derive the least squares estimator of μ_k for Model 1. SHOW ALL WORK!!!!

$$Q = \sum_i \sum_j (y_{ij} - \mu_i)^2$$

$$\frac{\partial Q}{\partial \mu_k} = -2 \sum_j (y_{kj} - \mu_k) \stackrel{\text{set}}{=} 0$$

$$\Rightarrow \sum_j y_{kj} = r_k \hat{\mu}_k \Rightarrow \hat{\mu}_k = \bar{y}_k.$$

p.5.b. Derive the least squares estimator of μ for Model 2. SHOW ALL WORK!!!!

$$Q = \sum_i \sum_j (y_{ij} - \mu)^2$$

$$\frac{\partial Q}{\partial \mu} = -2 \sum_i \sum_j (y_{ij} - \mu) \stackrel{\text{set}}{=} 0$$

$$\Rightarrow \sum_i \sum_j y_{ij} = \hat{\mu} \sum_i r_i \Rightarrow \hat{\mu} = \frac{\sum_i \sum_j y_{ij}}{\sum_i r_i} = \bar{y}_{..}$$

p.5.c. Three brands of cellphones are to be compared in terms of sound quality. Samples of $r_1 = 2$, $r_2 = 3$, $r_3 = 2$ are obtained, with the following sound quality scores: Brand 1: 8, 9; Brand 2: 8, 7, 6; Brand 3: 5, 6.

Brand1	Brand2	Brand3
8	8	5
9	7	6
	6	

Mod 1 8.5 7 5.5 $\hat{\mu}$
 Mod 2 7 7 7

Obtain the Error Sum of Squares and Degrees of Freedom for each model:

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 Model 2: $SSE_2 = (8-7)^2 + (9-7)^2 + (8-7)^2 + (7-7)^2 + (6-7)^2 + (5-7)^2 + (6-7)^2$ $df_{E2} = 7-1 = 6$
 $= 1 + 4 + 1 + 0 + 1 + 4 + 1 = 12$

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 Model 1: $SSE_1 =$ $df_{E1} = 7-3 = 4$
 $= (-0.5)^2 + (0.5)^2 + (1)^2 + 0^2 + (-1)^2 + (-0.5)^2 + (0.5)^2$
 $= 0.25 + 0.25 + 1 + 0 + 1 + 0.25 + 0.25 = 3$

p.5.d. Conduct the General Linear Test to test $H_0: \mu_1 = \mu_2 = \mu_3$

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 $F_{obs} = \frac{SSE_2 - SSE_1}{df_{E2} - df_{E1}} = \frac{12-3}{6-4} = \frac{9}{2} \left(\frac{4}{3}\right) = 6$
 $\frac{SSE_1}{df_{E1}} = \frac{3}{4}$

$F(.05; 2, 4) = 6.944$

Test Statistic: 6 Rejection Region: $F_{obs} \geq 6.944$

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