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I pledge my honor that I have abided by the Stevens Honor System.

## Query 2

### Problem 1

$$\bar{X}_{\cdot,} = \frac{1}{n_1 + \dots + n_k} \sum_{i=1}^k \sum_{j=1}^{n_i} X_{i,j} = \frac{n_1 \bar{X}_{1,\cdot} + \dots + n_k \bar{X}_{k,\cdot}}{n_1 + \dots + n_k}$$

From slides 7:

$$\bar{X}_{\cdot,} = \frac{1}{n} \sum_{i=1}^k \sum_{j=1}^{n_i} X_{i,j} = \frac{1}{n} \sum_{i=1}^k n_i \bar{X}_{i,\cdot}$$
$$\Rightarrow \sum_{i=1}^k \sum_{j=1}^{n_i} X_{i,j} = \sum_{i=1}^k n_i \bar{X}_{i,\cdot}$$
$$\frac{1}{n_1 + \dots + n_k} \cdot \left[ \sum_{i=1}^k \sum_{j=1}^{n_i} X_{i,j} \right] = \frac{1}{n_1 + \dots + n_k} \cdot \left[ n_1 \bar{X}_{1,\cdot} + \dots + n_k \bar{X}_{k,\cdot} \right]$$
$$\sum_{i=1}^k \sum_{j=1}^{n_i} X_{i,j} = n_1 \bar{X}_{1,\cdot} + \dots + n_k \bar{X}_{k,\cdot}$$
$$\sum_{i=1}^k n_i \bar{X}_{i,\cdot} = n_1 \bar{X}_{1,\cdot} + \dots + n_k \bar{X}_{k,\cdot}$$
$$n_1 \bar{X}_{1,\cdot} + \dots + n_k \bar{X}_{k,\cdot} = n_1 \bar{X}_{1,\cdot} + \dots + n_k \bar{X}_{k,\cdot}$$

### Problem 2

$$SSB = \sum_{i=1}^{\overbrace{k}^{\text{\# groups}}} \sum_{j=1}^{n_i} (\bar{X}_{i,\cdot} - \bar{X}_{\cdot,})^2 = \sum_{i=1}^k n_i (\bar{X}_{i,\cdot} - \bar{X}_{\cdot,})^2$$

in each group, the # of observations is constant

Each group's sum of squares is just

$$\sum_{j=1}^{n_i} (\bar{X}_{i,\cdot} - \bar{X}_{\cdot,})^2 = n_i (\bar{X}_{i,\cdot} - \bar{X}_{\cdot,})^2$$

To get the sum of squares between all k groups we get the summation using this

$$\sum_{i=1}^k n_i (\bar{X}_{i,\cdot} - \bar{X}_{\cdot,})^2$$

### Problem 3

$$SSE = \sum_{i=1}^k \sum_{j=1}^{n_i} (X_{i,j} - \bar{X}_{i,\cdot})^2 = \sum_{i=1}^k (n_i - 1) S_i^2.$$

From slides 7 (pg 20), Group sample variance

$$S_i^2 = \frac{1}{n_i - 1} \sum_{j=1}^{n_i} (X_{i,j} - \bar{X}_{i,\cdot})^2$$

we can cross multiply and see that

$$S_i^2 (n_i - 1) = \sum_{j=1}^{n_i} (X_{i,j} - \bar{X}_{i,\cdot})^2$$

So we can make this substitution into the LHS:

$$SSE = \sum_{i=1}^k (n_i - 1) S_i^2$$

### Problem 4

According to lecture 7,  $\psi = a_1 \mu_1 + \dots + a_k \mu_k$ .

### Problem 5

The lecture 7 notes give us the following:  $\psi = a_1 \mu_1 + \dots + a_k \mu_k$ , which we can rewrite as

$\psi = \sum_{i=1}^k a_i \mu_i$ . The notes also tell us  $C_\psi = \sum_{i=1}^k a_i \bar{X}_{i,\cdot}$ , and we can approximate  $\psi$  as  $E[C_\psi]$ . We can

simplify this knowing that  $\bar{X}_{i,\cdot} \sim N(\mu_i, \sigma^2/n_i)$ ,  $i = 1, \dots, k$  are independent.

$$E[C_\psi] = E\left[\sum_{i=1}^k a_i \bar{X}_{i,\cdot}\right] = \sum_{i=1}^k a_i E[\bar{X}_{i,\cdot}] = \sum_{i=1}^k a_i \mu_i = \psi.$$

### Problem 6

According to lecture 7:

$$\text{Var}[C_\psi] = \text{Var}\left[\sum_{i=1}^k a_i \bar{X}_{i,\cdot}\right] = \sum_{i=1}^k \text{Var}[a_i \bar{X}_{i,\cdot}] = \sum_{i=1}^k a_i^2 \text{Var}[\bar{X}_{i,\cdot}] = \sigma^2 \sum_{i=1}^k a_i^2 / n \equiv \sigma_\psi^2.$$

This is possible because  $\bar{X}_{i,\cdot} \sim N(\mu_i, \sigma^2/n_i)$ ,  $i = 1, \dots, k$  are independent.

### Problem 7

$C_\psi$  is a linear combination of standard normal random variables, therefore  $C_\psi \sim N(\psi, \sigma_\psi^2)$

### Problem 8

i

Under  $H_0: \psi = 0$ ,  $C_\psi$  follows a standard normal distribution. Therefore

$$P_{H_0}(|C_\psi| > 3) = 2(1 - \text{pnorm}(3)) = 0.002699796.$$

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> 2*(1-pnorm(3))
```

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[1] 0.002699796
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ii

Under  $H_a: \psi > 0$ , we must assume that  $C_\psi > 0$ . This is because  $C_\psi$  follows a normal distribution, so it is more likely that, even if we subtract 0.5,  $C_\psi > 0$ . Therefore

$$P(C_\psi > 0) - \frac{1}{2} > 0, \text{ thus its sign is positive.}$$