## 1 growth

We fit a second-order model using the following equation, check the summary and analysis of variance.

```
> h<-lm(Yield~x1+x2+x3+I(x1^2)+I(x2^2)+I(x3^2)+x1*x2+x2*x3+x1*x3,data = growth)
> summary(h)
> pure.error.anova(h)
```

The summay and anova table are shown as shown in figure 1 and figure 2

```
> summary(h)
call:
lm(formula = Yield \sim x1 + x2 + x3 + I(x1^2) + I(x2^2) + I(x3^3)
     x1 * x2 + x2 * x3 + x1 * x3, data = growth)
Residuals:
Min 1Q Median 3Q Max
-15.6661 -9.1577 -0.6661 9.1718 17.3339
Coefficients: Estimate Std. Error t value Pr(>|t|)
                                                     5.7e-09 ***
0.73765
(Intercept) 100.666
                                  5.564 18.093
3.691 0.344
                   1.271
x2
                                   3.691
                                             0.369
x3
I(x1^2)
                   -1.494
-3.767
                                  3.691
3.593
                                           -0.405
                                                      0.69411
                                           -1.048
                                                      0.31912
I(x2^2)
I(x3^2)
                                           -3.459
-2.672
                 -12.430
                                  3.593
                                                      0.00613
                   -9.601
                                  3.593
                                                      0.02342
                                           0.596
x1:x2
                   2.875
                                  4.823
                                                      0.56436
                   -4.625
                                  4.823
                                                      0.36020
x2:x3
x1:x3
                   -2.625
                                  4.823
                                           -0.544
---
Signif. codes:
0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 13.64 on 10 degrees of freedom
Multiple R-squared: 0.6631, Adjusted R-squared: 0.
F-statistic: 2.186 on 9 and 10 DF, p-value: 0.1194
```

Figure 1: summary of second order model

```
> pure.error.anova(h)
Analysis of Variance Table
Response: Yield
                         Sum Sq Mean Sq F value Pr(>F)
22.08 22.08 0.1285 0.73467
                   Df
x1
x2
                          25.31
30.50
                                       25.31
                                                 0.1473 0.71695
                                       30.50
                                                  0.1775 0.69105
х3
I(x1^2)
                                       46.37
                                                  0.2698 0.62564
                    1 1916.92 1916.92 11.1535 0.02056
1 1328.46 1328.46 7.7296 0.03889
I(x2^2)
I(x3^2)
x1:x2
x2:x3
                        66.12
171.13
                                       66.12 0.3847 0.56225
                                     171.13
                                                  0.9957 0.36417
X::X3 1 55.13 55.13
Residuals 10 1860.95 186.09
Lack of fit 5 1001.61 200.32
Pure Error 5 859.33 171.87
x1:x3
Residuals
                                     55.13
186.09
                                                 0.3207 0.59564
                                      200.32
                                                 1.1656 0.43529
```

Figure 2: anova of second order model

The p-value of lack of fit is 0.43529, which is greater than 0.05. Thus, we could conclude that the second model is adequate to represent the data. Fitted model:  $\hat{y} = 100.666 + 1.271\hat{x}_1 + 1.361\hat{x}_2 - 1.494\hat{x}_3 - 3.767\hat{x}_1^2 -$ Fitted model:  $y = 100.000 + 1.271x_1 + 1.501x_2 - 1.454x_3 - 5.701x_1$   $12.430\hat{x}_2^2 - 9.601\hat{x}_3^2 + 2.875\hat{x}_1\hat{x}_2 - 4.625\hat{x}_2\hat{x}_3 - 2.625\hat{x}_1\hat{x}_3$   $B_{3,3} = \begin{pmatrix} -3.7670 & 1.4375 & -1.3125 \\ 1.4375 & -12.4300 & -2.3125 \\ -1.3125 & -2.3125 & -9.6010 \end{pmatrix}, b = \begin{pmatrix} 1.271 \\ 1.361 \\ -1.494 \end{pmatrix}, x_s = -\frac{1}{2}B^{-1}b = \begin{pmatrix} 0.260 \\ 0.111 \\ -0.140 \end{pmatrix}, \text{ and the eigenvalues of matrix B are } \begin{pmatrix} -3.078495 \\ -8.953226 \\ -13.766279 \end{pmatrix}. \text{ All eigen-}$ 

$$\begin{pmatrix} 0.260 \\ 0.111 \\ -0.140 \end{pmatrix}$$
, and the eigenvalues of matrix B are  $\begin{pmatrix} -3.078495 \\ -8.953226 \\ -13.766279 \end{pmatrix}$ . All eigen-

values are negative, which makes sure that  $X_s$  is the maximum point.

## $\mathbf{2}$ average age

a) The sample design is simple random sampling without replacement. Under SRSWOR, the sample mean  $\bar{y}$  is an unbiased estimator of Y, thus the estimator of mean age for children is  $\bar{y} = \frac{9*13+10*35+11*44+12*69+13*36+14*24+15*7+16*3+17*2+18*5}{240} = 12.08$  The  $v(\bar{y})$  is an unbiased estimator of  $V(\bar{y})$ , and  $v(\bar{y}) = \frac{s^2}{n} = \frac{3.705}{240} = 0.015$ , thus, the standard error  $se(\bar{y}) = \sqrt{v(\bar{y})} = 0.124$ . And the 95% confidence interval for the average age is  $\bar{y}\pm Z_{\alpha/2}s\sqrt{\frac{1}{n}}=12.08\pm0.243$  b) We determine the sample size based on this formula:  $n=\frac{Z_{\alpha/2}^2S^2}{e^2}=\frac{1.96^2*3.705}{0.5^2}=$ 56.93, hence, the minimum sample size is 57.

## 3 clams

First, we calculate  $N_h h = 1, ..., 4, N_1 = 222.81 * 25.6 = 5704, N_2 = 49.61 *$  $25.6 \, = \, 1270, N_3 \, = \, 50.25 * 25.6 \, = \, 1287, N_4 \, = \, 197.81 * 25.6 \, = \, 5064, N \, = \, 1287 + 12$  $N_1 + N_2 + N_3 + N_4 = 13325$ . Then we obtain the  $\bar{y_{st}} = \sum_{h=1}^{H} W_h \bar{y_h} = 1.36$ . After that, we can have the estimator of the total number of bushels  $\hat{t_{st}} = \sum_{h=1}^{H} W_h \bar{y_h} = 1.36$ .  $N\bar{y_{st}} = 13325 * 1.36 = 18122.$ 

The variance of  $\hat{y_{st}}$ :  $v(\hat{y_{st}}) = \sum_{h=1}^{H} W_h^2 (1 - n_h/N_h) s_h^2/n_h = 0.0327$ , thus, the variance of  $\hat{t_{st}}$ :  $v(\hat{t_{st}}) = N^2 v(y_{st}) = 13325^2 * 0.0327 = 5806069$  the standard error is  $se(\hat{t_{st}}) = \sqrt{v(\hat{t_{st}})} = 2410$ 

## totoal number of acres 4

a) Use ratio estimation to estimate the total number of acres:

$$\hat{R} = \frac{\bar{y}}{\bar{x}} = \frac{mean(acres92)}{mean(farms87)} = 459.8975$$

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$$\hat{t_{yr}} = \hat{R}t_x = 459.8975 * 2087759 = 960, 155, 061$$

**b)** Use the regression estimation:

$$\hat{\beta}_0 = 263098.45, \hat{\beta}_1 = 58.09$$

$$\hat{\bar{y}}_{req} = \hat{\beta}_0 + \hat{\beta}_1 \bar{x} = 263098.45 + 58.09 * 2087759/3078 = 302500$$

$$\hat{t}_{yreq} = N\hat{\bar{y}}_{req} = 3078 * 302500 = 931,095,000$$

c) In order to find the method with most precision, we calculate the standard variance of  $\hat{t}_y$ . ratio estimation with auxiliary variable acres 87,

$$se(\hat{t}_{yra87}) = \sqrt{var(\hat{t}_y)} = \sqrt{N^2(1 - \frac{n}{N})\frac{1}{n}\frac{1}{n-1}\sum_{i \in s}(y_i - \hat{R}x_i)^2} = 5,344,567$$
 ratio estimation with auxiliary variable farms87,

$$se(\hat{t}_{yrf87}) = \sqrt{var(\hat{t}_y)} = \sqrt{N^2(1 - \frac{n}{N})\frac{1}{n}\frac{1}{n-1}\sum_{i \in s} (y_i - \hat{R}x_i)^2} = 65,364,822$$
 regression estimation with auxiliary variable farms87,

$$se(\hat{t}_{yregf87}) = \sqrt{var(\hat{t}_y)} = \sqrt{N^2(1 - \frac{n}{N})\frac{1}{n}\frac{1}{n-1}\sum_{i \in s} (y_i - \beta_0 - \beta_1 * x_i)^2} = 58,065,813$$