

Section 1

MANE 6313

Subsection 1

Week 12, Module **B**

Student Learning Outcome

- Select an appropriate experimental design with one or more factors,
- Select an appropriate model with one or more factors,
- Evaluate statistical analyses of experimental designs,
- Assess the model adequacy of any experimental design, and
- Interpret model results.

Module Learning Outcome

Assessing linear regression assumptions.

Fitting Linear Regression Models

Model Assumptions and Residuals

- Least squares estimation requires that $E(\varepsilon) = 0$ and $V(\varepsilon) = \sigma^2$ and the $\{\varepsilon_i\}$ are uncorrelated
- To perform statistical hypothesis tests, we further assume that $\varepsilon \sim \text{NID}(0, \sigma^2)$
- These assumptions are validated by examining the residuals

Test for Significance of Regression

- Test for significance of regression is a test to determine if there is a linear relationship between y and a subset of the regressors

$$H_0 : \beta_1 = \beta_2 = \cdots = \beta_k = 0$$

$$H_a : \beta_j \neq 0 \text{ for at least one } j$$

- The test statistic is

$$F_0 = \frac{SS_R/k}{SS_E/(n-k-1)} = \frac{MS_R}{MS_E}$$

- Reject H_0 if $F_0 > F_{\alpha, k, n-k-1}$

Example 12.8

```

27 > ```{r}
28 ex12_8.model <- lm(GrainRadius~PowderTemp+Extrusion+DieTemp,data=ex12_8.df)
29 summary(ex12_8.model)
30 ^ ```

```

Call:

```
lm(formula = GrainRadius ~ PowderTemp + Extrusion + DieTemp,
    data = ex12_8.df)
```

Residuals:

1	2	3	4	5	6	7	8
-0.75	5.25	1.75	-2.25	-6.25	-2.25	1.25	3.25

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-284.50000	30.77729	-9.244	0.000761 ***
PowderTemp	0.12500	0.08501	1.470	0.215398
Extrusion	2.45833	0.28336	8.676	0.000972 ***
DieTemp	1.45000	0.11335	12.793	0.000215 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 4.809 on 4 degrees of freedom

Multiple R-squared: 0.9837, Adjusted R-squared: 0.9714

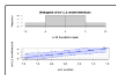
F-statistic: 80.36 on 3 and 4 DF, p-value: 0.0004967

Residuals - Normality Assumption

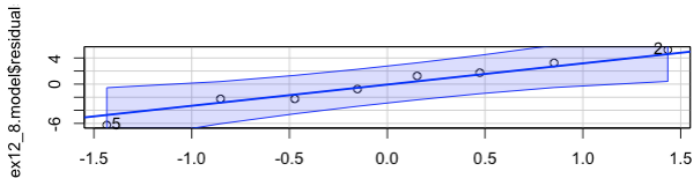
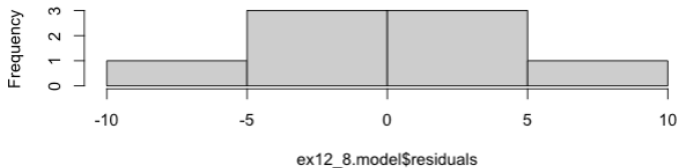
```

34 > ```{r}
35 par(mfrow=c(2,1))
36 hist(ex12_8.model$residuals)
37 library(car)
38 qqPlot(ex12_8.model$residuals)
39 > ```

```



Histogram of ex12_8.model\$residuals



Residuals vs.Fitted Values

```
54 > `[[r]]`  
55 ex12_8.stdres <- rstandard(ex12_8.model)  
56 plot(ex12_8.model$fitted.values,ex12_8.stdres,xlab="Fitted Values",ylab="Standardized Residuals",main="Residuals vs. Fitted Values")  
57 >
```

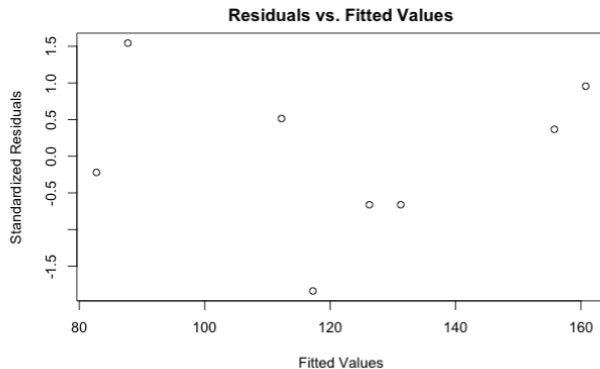


Figure 3: Ex. 12.8 Residuals vs. Fitted Values

Residuals vs. Powder Temperature

```
41 {r warning=FALSE,message=FALSE}  
42 ex12_8.stdres <- rstandard(ex12_8.model)  
43 plot(ex12_8.df$PowderTemp,ex12_8.stdres,xlab="Powder Temperature",ylab="Standardized Residuals",main="Residuals vs.  
44 Powder Temperature")
```

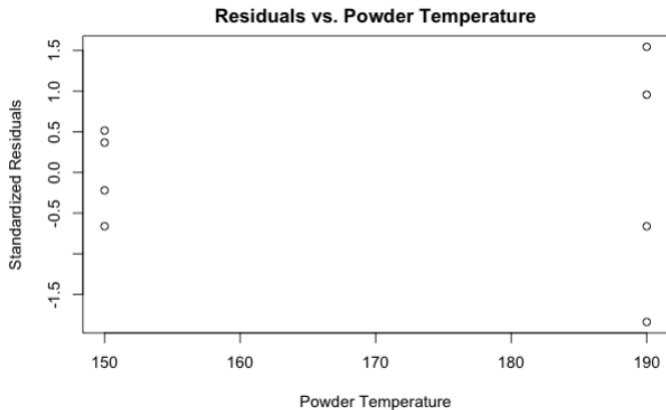


Figure 4: Residuals vs. Powder Temperature

Residuals vs. Extrusion

```
46 ~ {r warning=FALSE,message=FALSE}  
47 plot(ex12_8.df$Extrusion,ex12_8.stdres,xlab="Extrusion",ylab="Standardized Residuals",main="Residuals vs. Extrusion")  
48 ~
```

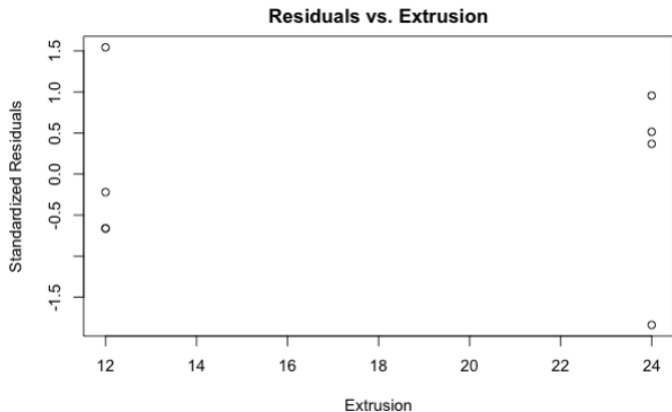


Figure 5: Residuals vs. Extrusion

Residuals vs. Die Temperature

```
50 {r warning=FALSE,message=FALSE}  
51 plot(ex12_8.df$DieTemp,ex12_8.stdres,xlab="Die Temperature",ylab="Standardized Residuals",main="Residuals vs. Die  
52 Temperature")
```

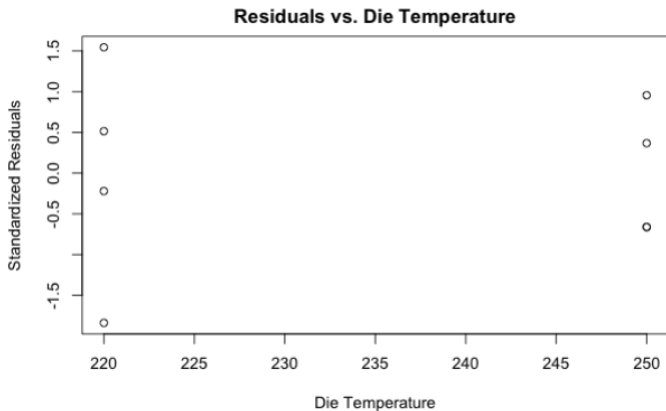


Figure 6: Residuals vs. Die Temperature