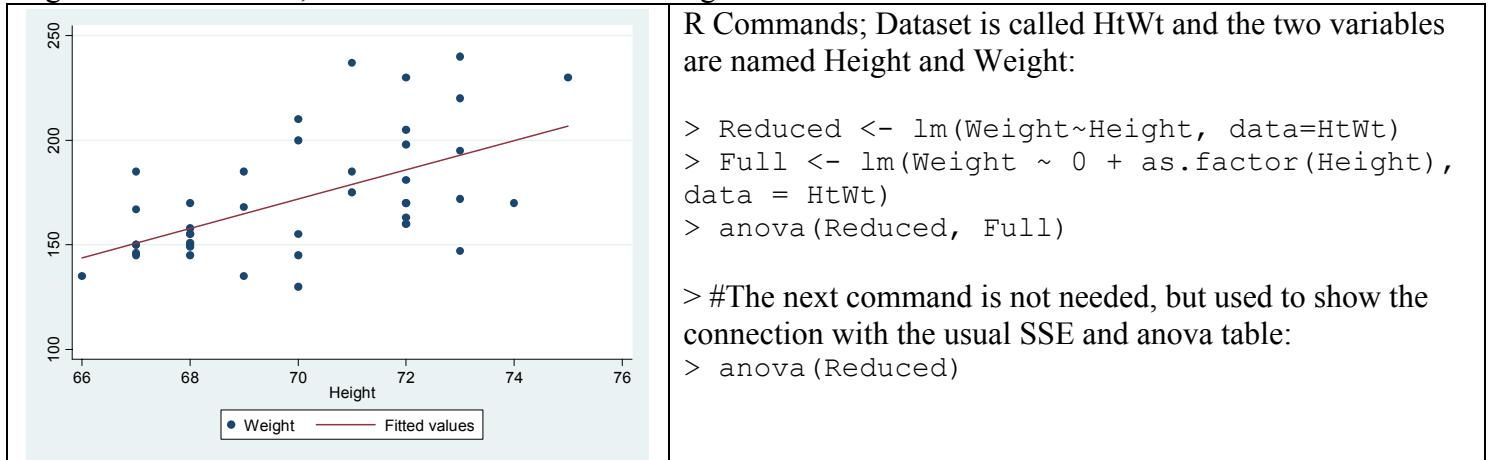


Lack of Fit Test Example Using Male Weight and Height Data

Data represent a sample of $n = 43$ college males measured at $c = 10$ different heights. There are multiple weight observations at most of the heights, which are measured to the nearest inch. Here is a plot of the data with the regression line shown, and the R Commands used to generate the test for lack of fit:



R Commands and Results, with lack of fit test results in bold:

```

> Reduced <- lm(Weight ~ Height, data = HtWt)
> Full <- lm(Weight ~ 0 + as.factor(Height), data = HtWt)
> anova(Reduced, Full)
Analysis of Variance Table

```

```

Model 1: Weight ~ Height                               [This is the reduced model]
Model 2: Weight ~ 0 + as.factor(Height)   [This is the full model]

```

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	41	23617				
2	33	20712	8	2905.1	0.5786	0.7879

```

> #The following is not needed, but shows the connection with regression.
> anova(Reduced) [This shows the anova table for the usual regression]

```

Analysis of Variance Table

Response: Weight

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
Height	1	11277	11277	19.578	6.978e-05 ***
Residuals	41	23617	576		

Notes:

- Model 1 is the usual linear regression model, which is the *reduced* model in this case; $SSE(R) = 23617$
- Model 2 is telling R to consider Height as a “Factor” instead of a continuous variable, thus treating it as categorical and fitting the mean at each height. $SSE(F) = 20712 = SSE(PE)$ [F = Full, PE = pure error]
- The Lack of Fit SSE is $SSE(LF) = SSE(R) - SSE(F) = 23617 - 20712 = 2905$
- $$F^* = \frac{2905.1}{\frac{8}{33}} = 0.5786$$
. Do not reject H_0 because the p -value = .7879.
- Conclude that using the linear regression is almost as good as using separate means at the 10 heights. Advantage of linear regression is that we can predict even for heights in between those measured.