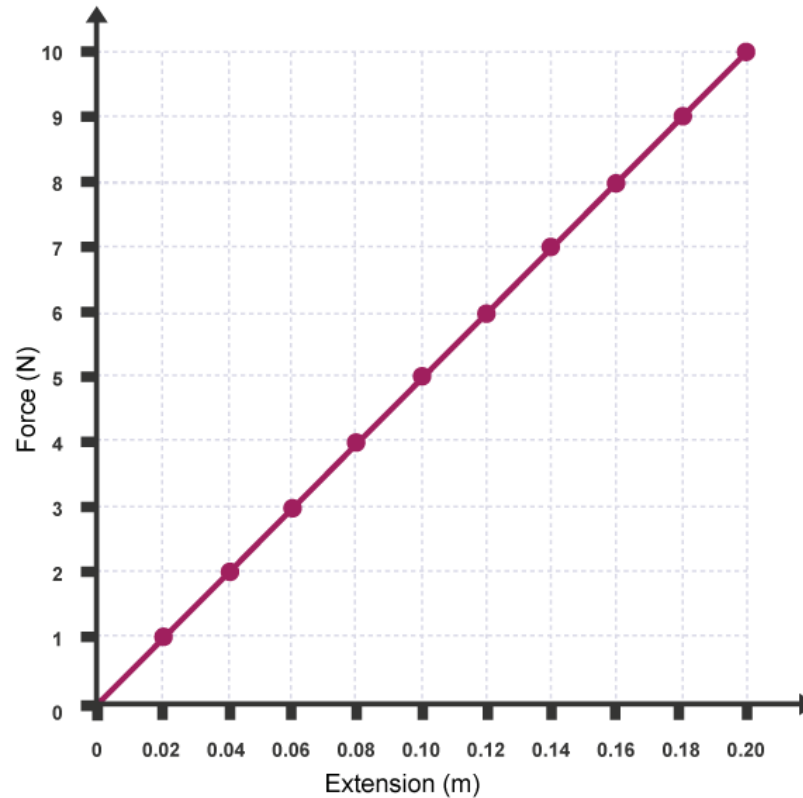
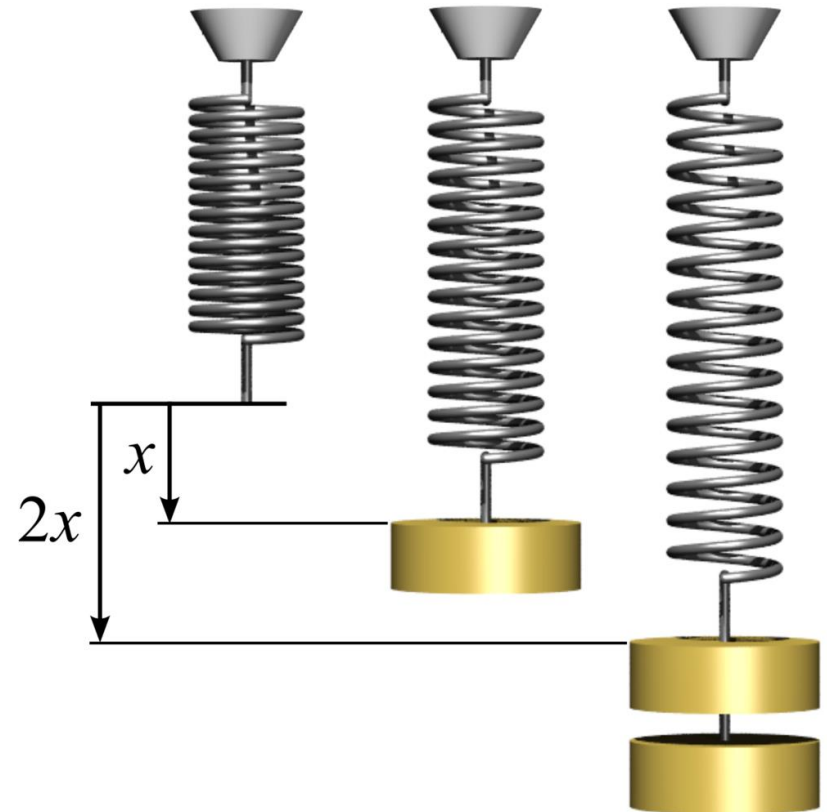


Fitting a Linear Regression with a Fixed Intercept



When Does it Make Sense to Use Zero Intercept?

- When you are sure that the intercept is zero.
- Example: Hooke's Law
 - The spring extends more for heavier weights
 - Doesn't extend at all for weight zero
 - $\Delta X = \beta_1 w$
 - $w=0$ must mean $\Delta X=0$



Qualitative Variables: A convenient way to estimate the mean

```
summary(lm(Percentage~0+Pair, data=fish))
```

```
##
## Call:
## lm(formula = Percentage ~ 0 + Pair, data = fish)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -52.429  -8.414   0.247  10.859  28.871
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## PairPair1     56.406     3.864   14.60  <2e-16 ***
## PairPair2     60.886     4.131   14.74  <2e-16 ***
## PairPair3     62.429     3.749   16.65  <2e-16 ***
## PairPair4     67.000     4.131   16.22  <2e-16 ***
## PairPair5     64.211     5.152   12.46  <2e-16 ***
## PairPair6     63.336     4.131   15.33  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 15.46 on 78 degrees of freedom
## Multiple R-squared:  0.9458, Adjusted R-squared:  0.9416
## F-statistic: 226.8 on 6 and 78 DF,  p-value: < 2.2e-16
```

Would have to add the value of the intercept to the coefficients to get the sample means if the intercept weren't 0

ANOVA when the intercept is 0

```
anova(lm(Percentage~0+Pair, data=fish))
```

```
## Analysis of Variance Table
##
## Response: Percentage
##           Df Sum Sq Mean Sq F value    Pr(>F)
## Pair         6 325175    54196  226.83 < 2.2e-16 ***
## Residuals   78  18637      239
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
anova(lm(Percentage~Pair, data=fish))
```

```
## Analysis of Variance Table
##
## Response: Percentage
##           Df Sum Sq Mean Sq F value Pr(>F)
## Pair         5   938.7   187.75  0.7858 0.563
## Residuals   78 18636.7   238.93
```

$$SSE = \sum_{ij} (y_{ij} - \hat{y}_i)^2$$

$$SST = \sum_{ij} (y_{ij} - 0)^2$$

$$SSR = SST - SSE = \sum_{ij} (\hat{y}_i - 0)^2$$

$$SSE = \sum_{ij} (y_{ij} - \hat{y}_i)^2$$

$$SST = \sum_{ij} (y_{ij} - \bar{y})^2$$

$$SSR = SST - SSE = \sum_{ij} (\hat{y}_i - \bar{y})^2$$

F-test

$$SSE = \sum_{ij} (y_{ij} - \hat{y}_i)^2$$

$$SST = \sum_{ij} (y_{ij} - 0)^2$$

$$SSR = SST - SSE = \sum_{ij} (\hat{y}_i - 0)^2$$

$$SSE = \sum_{ij} (y_{ij} - \hat{y}_i)^2$$

$$SST = \sum_{ij} (y_{ij} - \bar{y})^2$$

$$SSR = SST - SSE = \sum_{ij} (\hat{y}_i - \bar{y})^2$$

Null Hypothesis: $\mu_1 = \mu_2 = \dots = 0$

F-statistic: $\frac{MSR}{MSE}$

$MSR = SSR/df_{SSR}$ is potentially huge

Null Hypothesis: $\mu_1 = \mu_2 = \dots = 0$

F-statistic: $\frac{MSR}{MSE}$

A Fix

- Suppose we ran ANOVA fixing the intercept at 0
- What we want is to test the hypothesis that
Null Hypothesis: $\mu_1 = \mu_2 = \dots = \bar{y}$
- Compute the right SSR by hand!
- (in R)