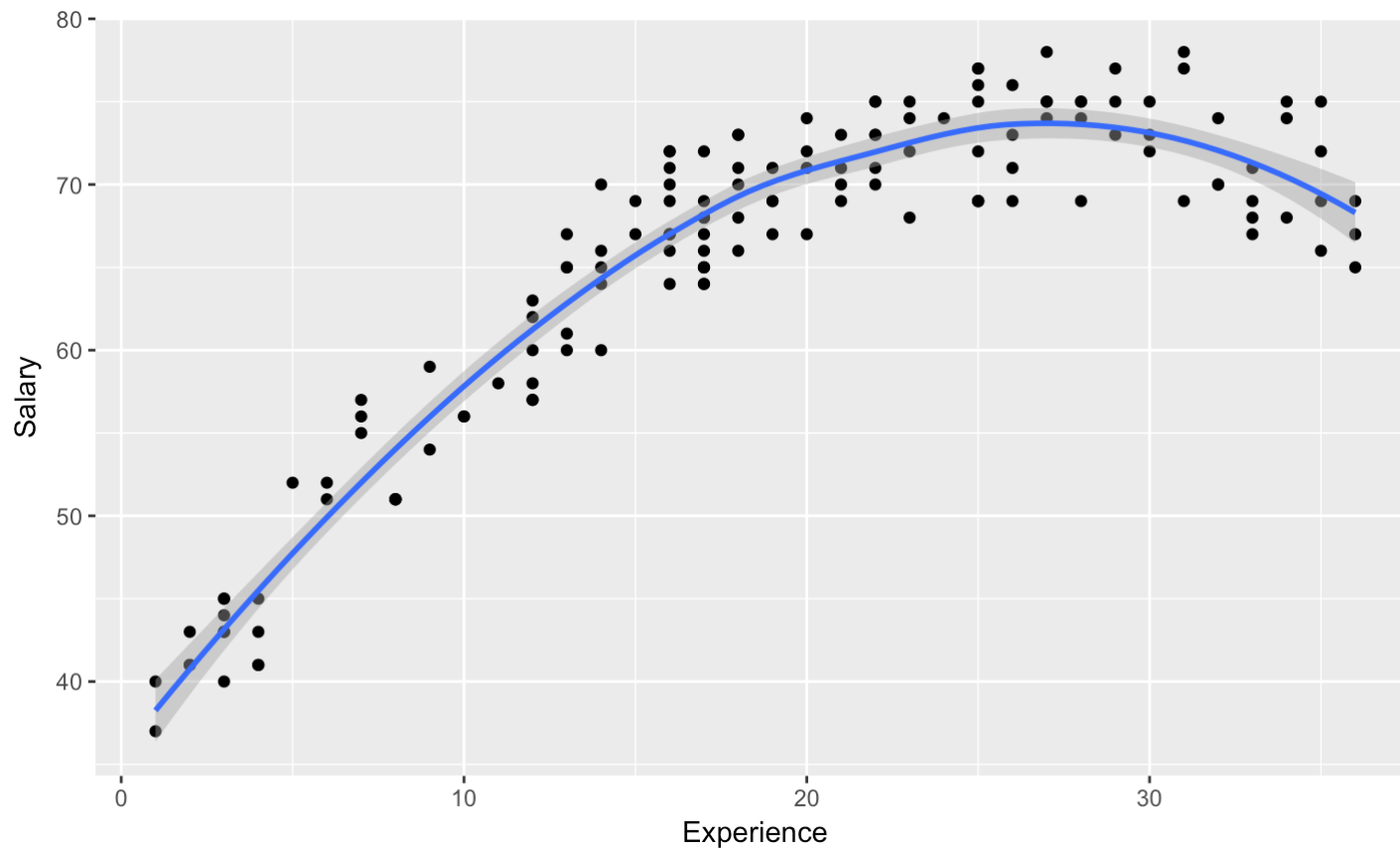


# Polynomial Regression

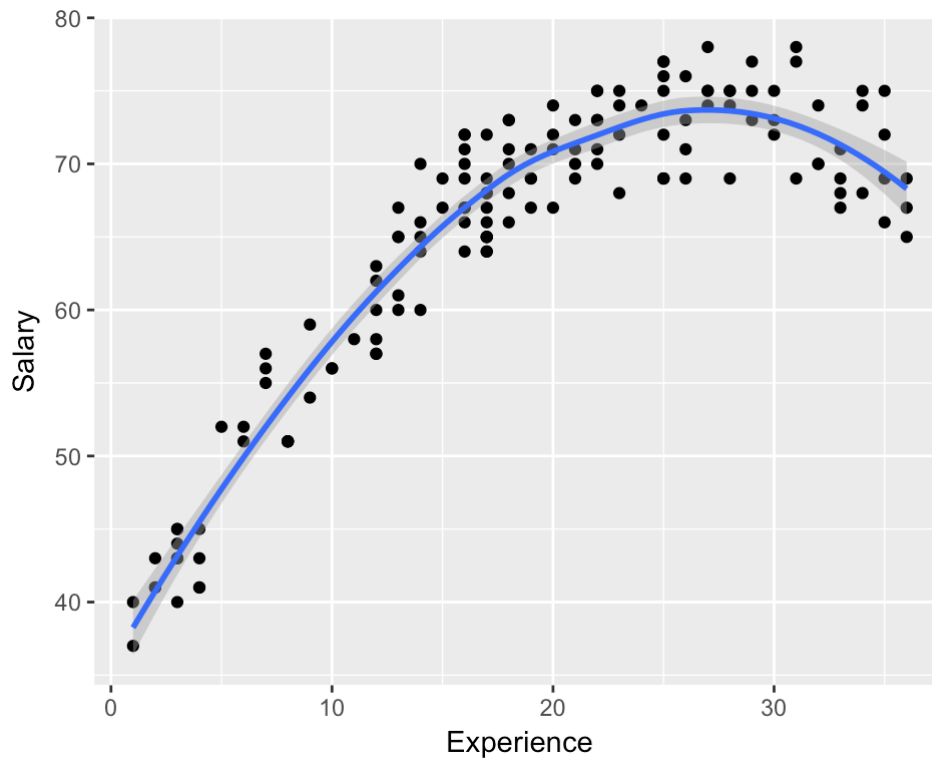
Math 430, Winter 2017

# Modeling salary

What if the relationship between salary and years of experience wasn't linear?



# Polynomial regression



A quadratic model seems sensible.

$$Y_i = \beta_0 + \beta_1 x_i + \beta_2 x_i^2 + e_i$$

# Polynomial regression in R

```
quad.mod <- lm(Salary ~ Experience + I(Experience^2), data = profsalary)
summary(quad.mod)

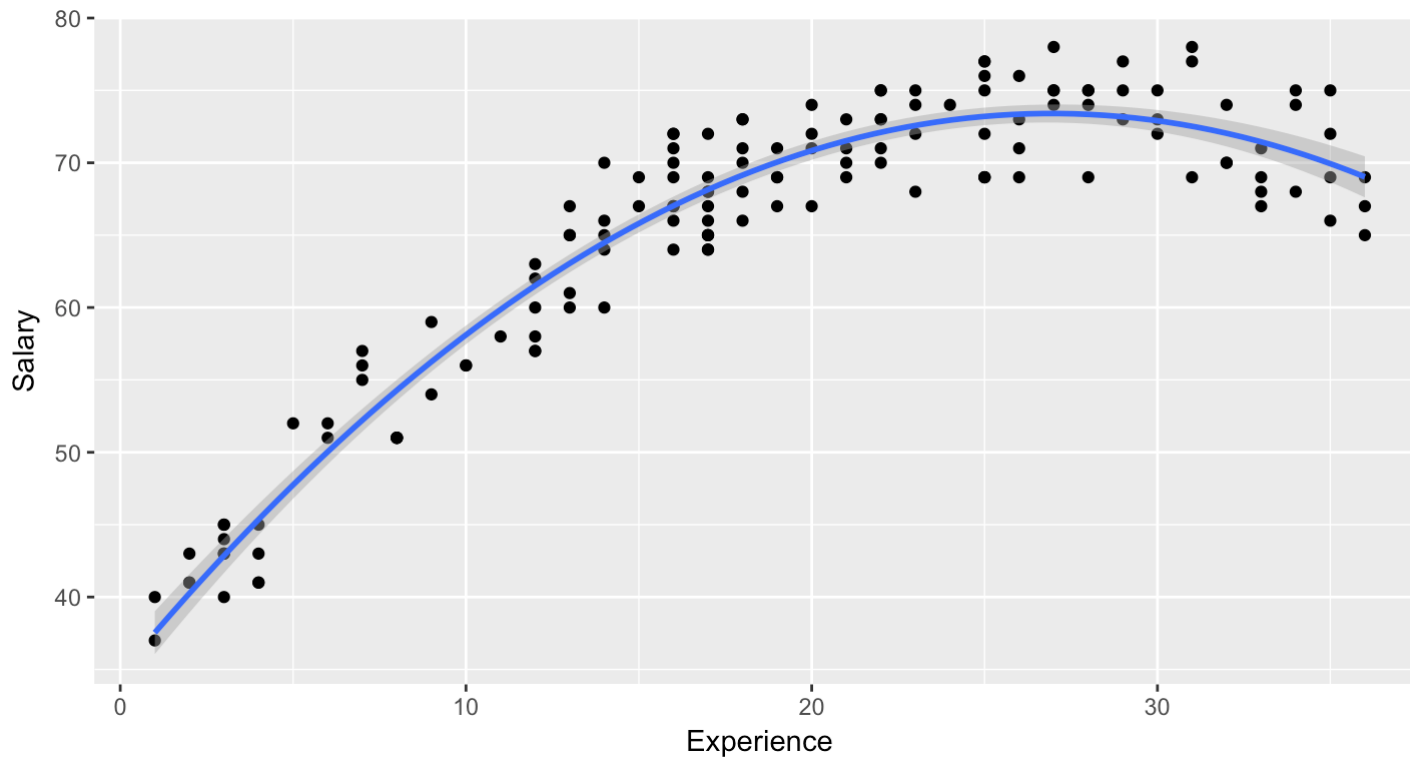
##
## Call:
## lm(formula = Salary ~ Experience + I(Experience^2), data = profsalary)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.5786 -2.3573  0.0957  2.0171  5.5176
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   34.720498   0.828724   41.90  <2e-16 ***
## Experience     2.872275   0.095697   30.01  <2e-16 ***
## I(Experience^2) -0.053316   0.002477  -21.53  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.817 on 140 degrees of freedom
## Multiple R-squared:  0.9247, Adjusted R-squared:  0.9236
## F-statistic: 859.3 on 2 and 140 DF, p-value: < 2.2e-16
```

# Interpreting coefficients

# Interpreting coefficients

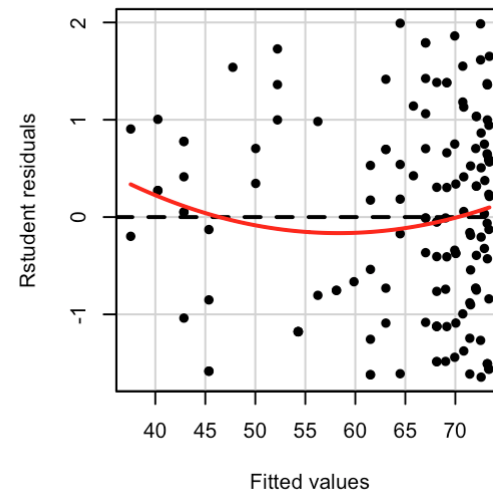
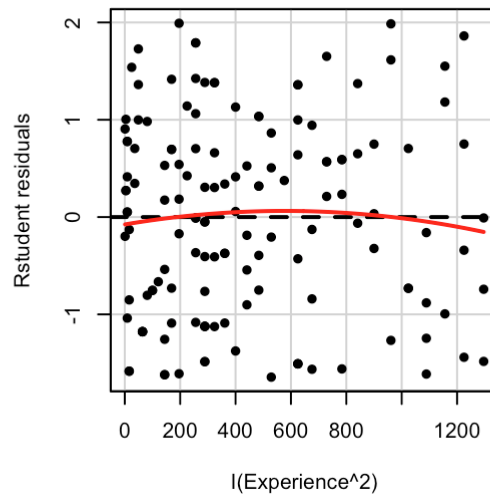
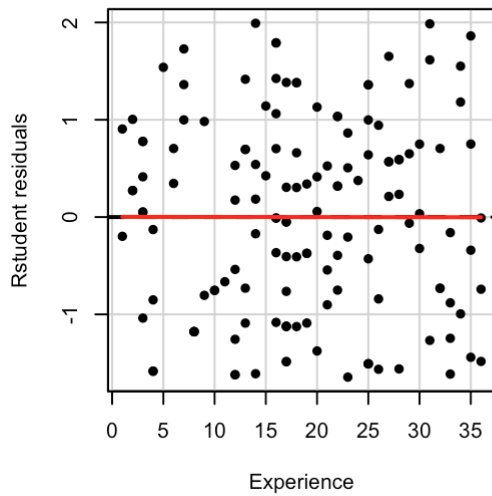
# Polynomial regression

```
ggplot(data = profsalary, aes(x = Experience, y = Salary)) +  
  geom_point() +  
  geom_smooth(method = "lm", formula = y ~ poly(x, 2))
```



# Model checking

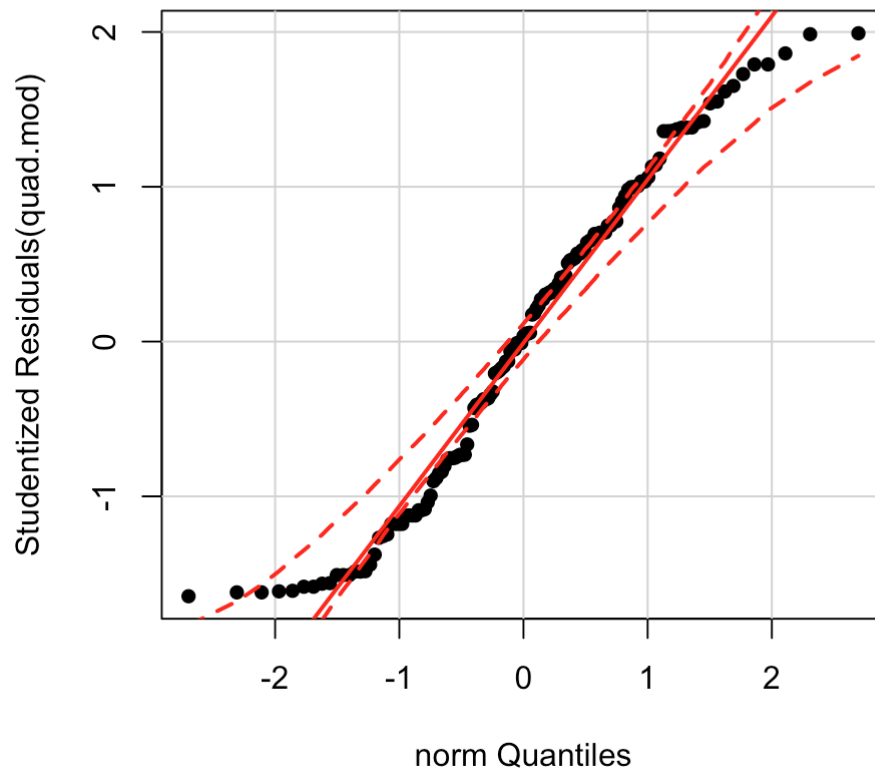
```
residualPlots(quad.mod, type = "rstudent", pch = 16)
```





# Model checking

```
qqPlot(quad.mod, pch = 16, dist = "norm", reps = 5000)
```



# Beyond second-order models

Polynomials for one predictor

$$Y = \beta_0 + \beta_1 x + \beta_2 x^2 + \cdots + \beta_p x^p + \varepsilon_i$$