

# Statistical Methods for Insurance: Linear Models

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W4.C2

# Overview of this class

- Fitting a linear model to olympic medal tally
- Review of linear regression
- READING: Ch 5, Diez, Barr, Cetinkaya-Rundel

# Modeling Olympic medal counts

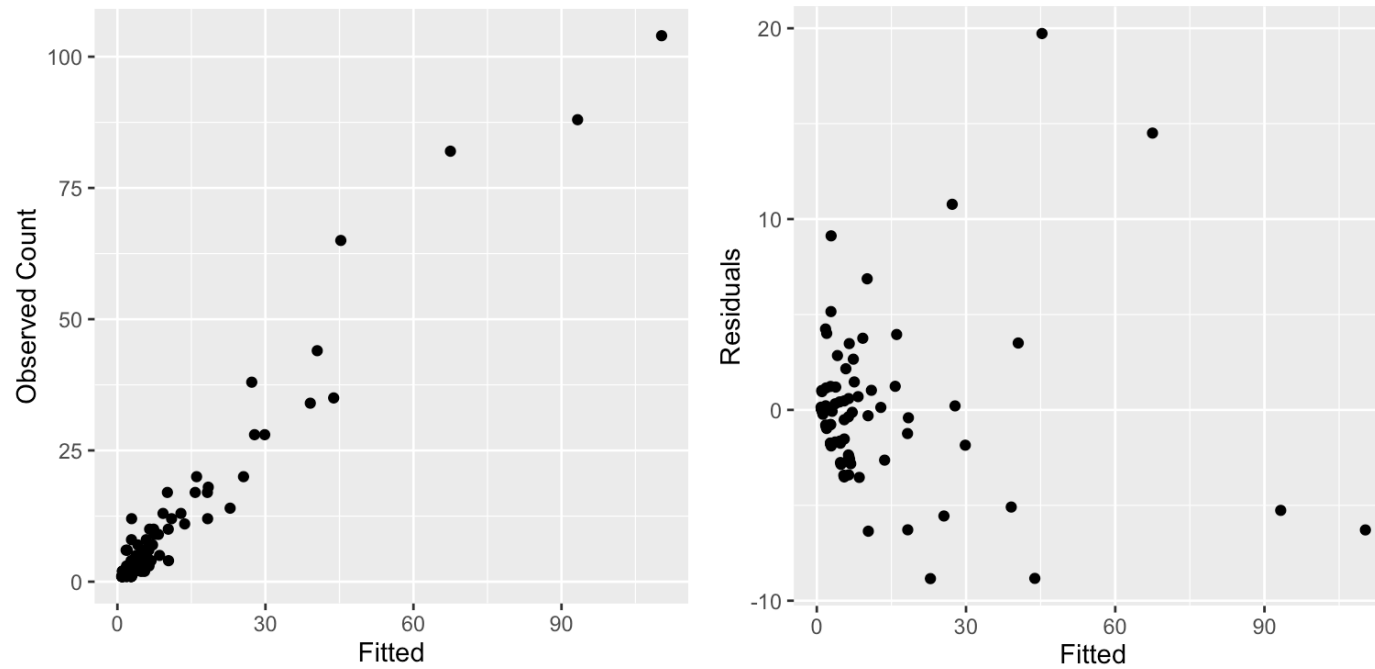
term	estimate	std.error	statistic	p.value
(Intercept)	1.28	0.67	1.91	0.06
Total.2008	0.91	0.04	20.80	0.00
Pop_std	-0.51	0.54	-0.94	0.35
GDP_std	1.27	0.85	1.50	0.14

$$M_{2012} = \beta_0 + \beta_1 M_{2008} + \beta_2 Population + \beta_3 GDP + \varepsilon$$

# Model summary

```
#>    null.deviance df.null logLik AIC BIC deviance df.residual
#> 1          30252      84  -242 494 507    1486          81
```

# Fit and residuals



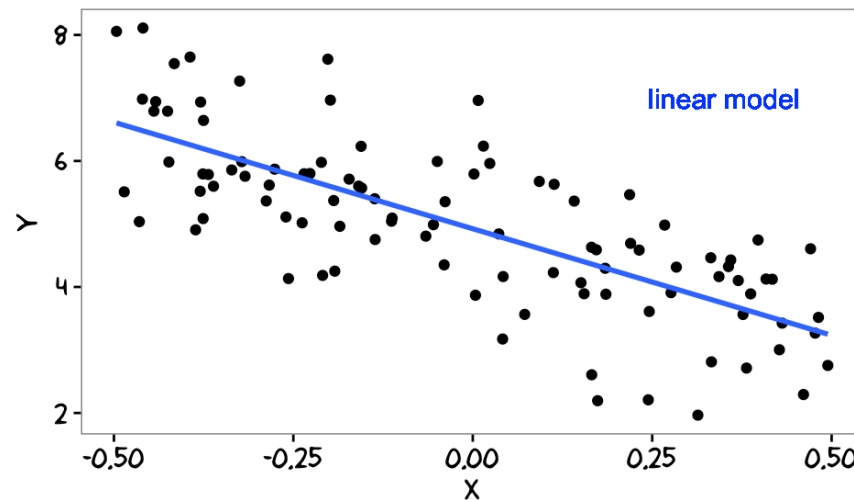
# Make plots interactive

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# Simple linear model

$$Y = \beta_0 + \beta_1 X + \varepsilon$$

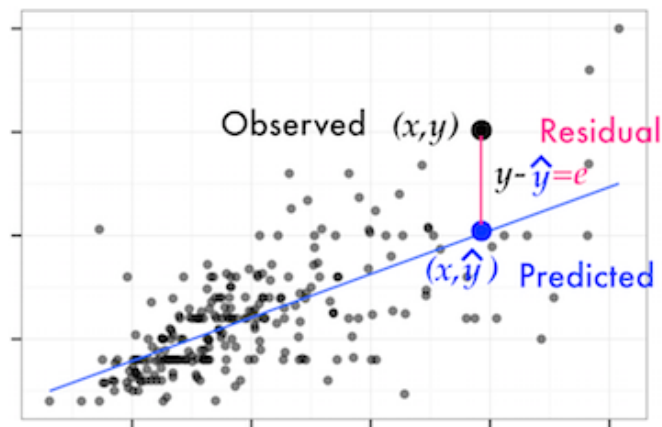
- Explains how response variable ( $Y$ ) changes in relation to explanatory variable ( $X$ ), on average.
- Use line to predict value of  $Y$  for a given value of  $X$





# Observed, fitted, residuals

- Observed value is  $Y$  (a point on plot)
- Fitted value is  $\hat{Y}$ , a value that lies on the line
- Residual is the difference between observed and fitted,  $e = Y - \hat{Y}$



# Fitting process

- Minimizing the sum of squared residuals produces the best fitting line.
- Minimizes  $\sum e^2$
- Line that is closest to the points, as a whole.

# Parameter interpretation

- Line of best fit:  $\hat{Y} = b_0 + b_1X$
- $b_0$  is the intercept of the line with y-axis
- $b_1$  is the slope of the line

# Calculating manually

Given standard deviation of  $X$ ,  $s_x$ , standard deviation of  $Y$ ,  $s_y$ , and the correlation,  $r$ , between the two, the slope is computed by

$$b_1 = r \frac{s_y}{s_x}$$

and given the sample means  $\bar{X}$ ,  $\bar{Y}$

$$b_0 = \bar{Y} - b_1 \bar{X}$$

# YOUR TURN

Is the point  $\bar{X}, \bar{Y}$  on the regression line?

# Prediction

For given  $X$  values, plug these into the model equation to predict  $Y$ ,

$$\hat{Y} = b_0 + b_1X$$

# Goodness of fit

- $R^2$  is the proportion of variation in  $Y$  that is explained by  $X$ . Computed by

$$R^2 = 1 - \frac{\sum e^2}{\sum Y^2}$$

- Deviance: up to a constant, minus twice the maximized log-likelihood

# Reading residual plots

- Make a histogram and normal probability plot of the residuals - for a good fit the shape should be pretty symmetric and bell-shaped
- Plot the residuals against the fitted values - for a good fit should be just a random splatter, no patterns

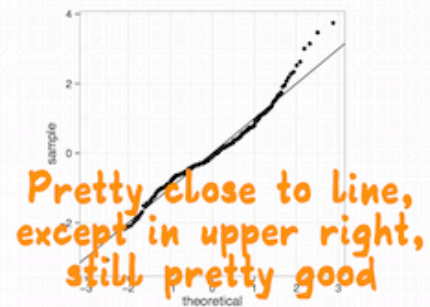


# Residual plots

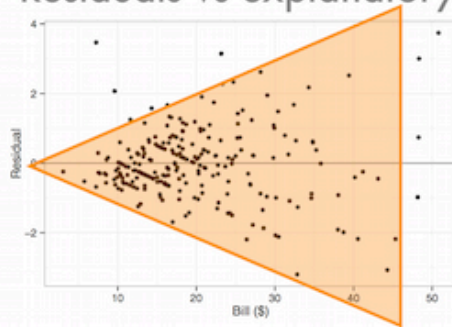
Histogram of residuals



Normal probability plot



Residuals vs explanatory variable



Plot exhibits heteroskedasticity, suggests that tip variability depends size of the bill.

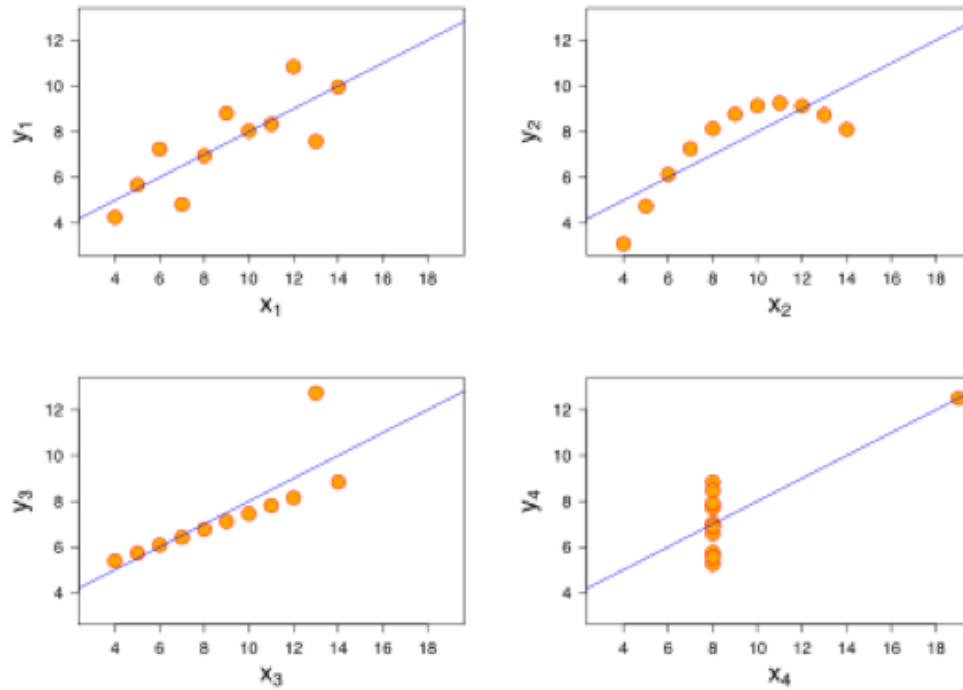
# Diagnostics

- Influential points: leverage (diagonal elements of hat matrix, values  $> 2p/n$  would indicate cases with high influence), cooksd (Cooks distance, measures the change in the residual when the case is removed)
- Collinearity between explanatory variables (multiple regression): variance inflation factor

# Cautions

- Association is not causation
- Linear association only
- Extrapolation outside the range of the data is not recommended

# Anscombe's quartet



Always plot the data, because very different patterns can lead to the same correlation.

# Resources

- [Statistics online textbook, Diez, Barr, Cetinkaya-Rundel](#)
- [Ancombe's quartet](#)

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