

## Lecture 34

(1)

### Residuals

Recall that the residual is the difference between the observed value of the response variable and the predicted value  $\hat{y}$

$$e_i = y_i - \hat{y}_i$$

It turns out that because residuals show how far the observed data falls from the regression line they are useful for assessing how well a fitted regression line ~~is~~ modelled the observed data. Note also that the mean of the residuals is always zero.

Specifically, the tools we use are called a residual plots.

# Residuals plot

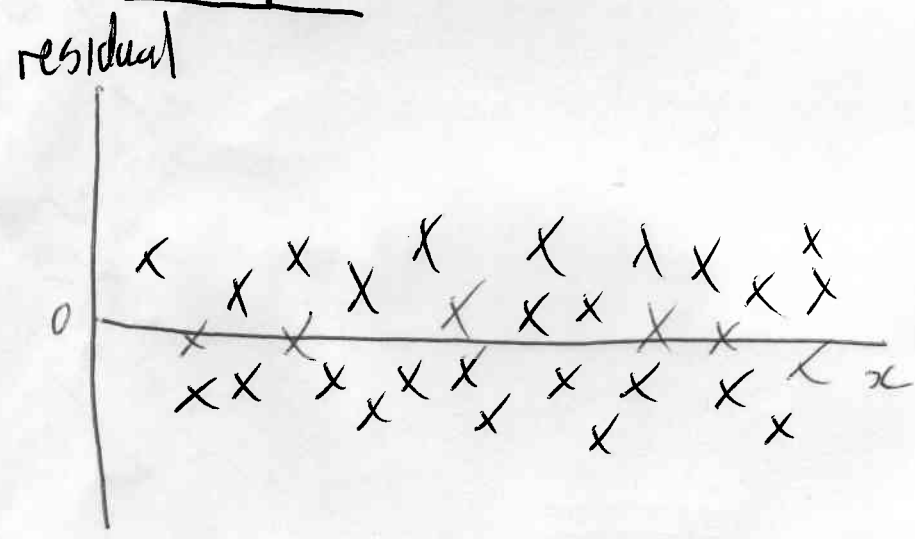
These are plots of residuals against \_\_\_\_\_

- residuals vs explanatory variable
- residuals vs predicted values

Note never plot residuals versus response variable.

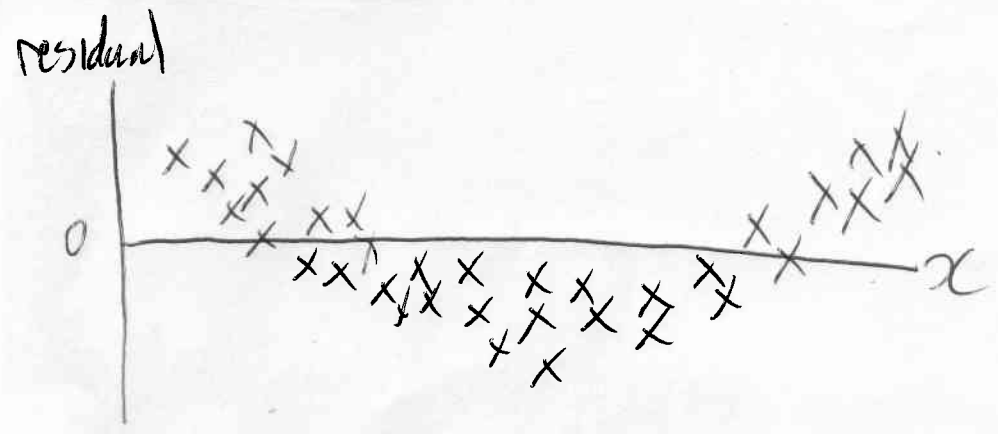
They are used to visually assess the fit of the regression line. ~~plot~~

## Ideal plot

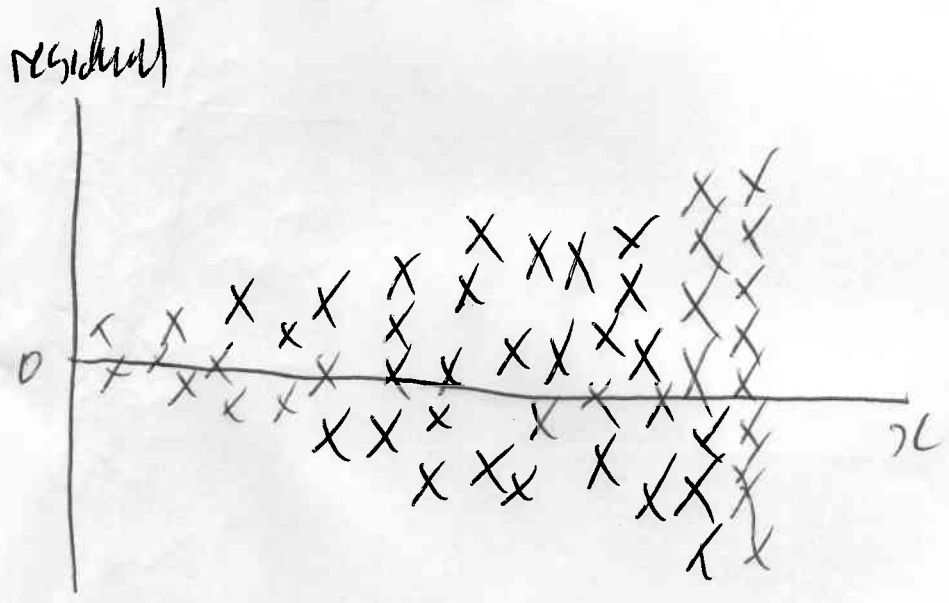


- Shows irregular pattern i.e. no clear pattern.
- Even <sup>vertical</sup> spread across all values on x axis

# Non ideal plots



- pattern (this particular pattern indicates a curved relationship exists between y and x that is not modelled by the <sup>linear</sup> regression line)



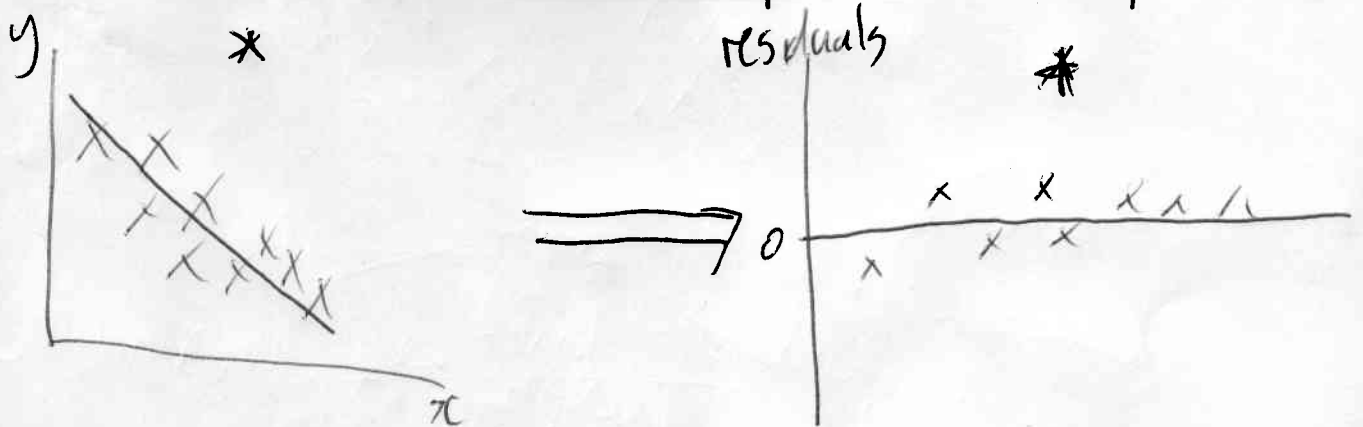
- An even vertical spread (this particular relationship indicates a transformation of y variable might be useful)

# Outliers/Influential observations

An outlier is an observation that lies outside the overall pattern of the other observations.

An observation is influential if removing it would have large effect on the fitted regression line (perhaps even completely changing the equation of the regression line)

— Outliers in  $y$  tend to show up in residuals plots as



- outliers in  $x$  are often (but not always) influential

eg

