

# M8S2 - Regression In Practice

Professor Jarad Niemi

STAT 226 - Iowa State University

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# Outline

## 1. Assumptions

- Independence
- Normality
- Constant variance
- Linearity

## 2. Regression analysis steps

- a. Determine scientific questions, i.e. why are you collecting data
- b. Collect data (at least two variables per individual)
- c. Identify explanatory and response variables
- d. Plot the data
- e. Run regression
- f. Assess regression assumptions
- g. Interpret regression output

# Regression assumptions

Regression model

$$y_i = \beta_0 + \beta_1 x_i + \epsilon_i \quad \epsilon_i \stackrel{iid}{\sim} N(0, \sigma^2)$$

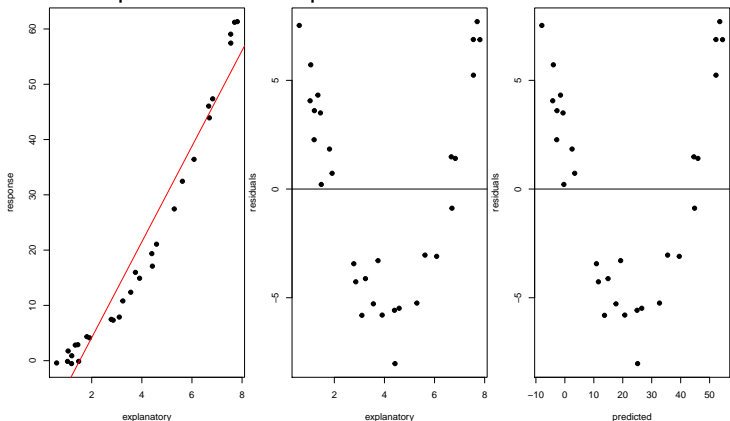
Regression assumptions are

- Errors are **independent**
- Errors are **normally distributed**
- Errors are identically distributed with a mean of 0 and **constant variance** of  $\sigma^2$
- Linear relationship between explanatory variable and mean of the response

# Assessing linearity assumption

Look for non-linearity in

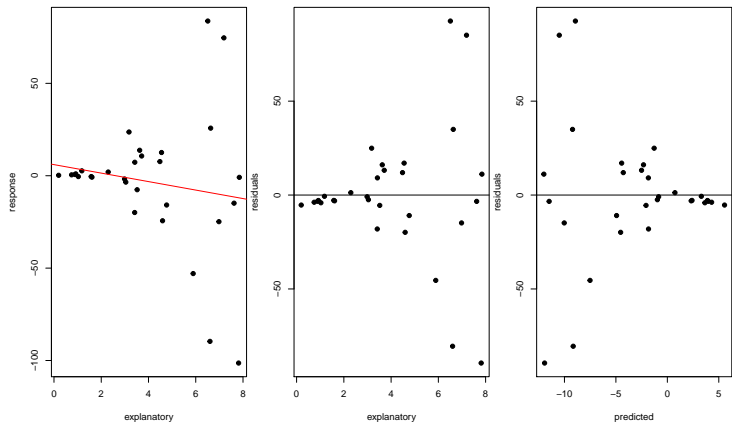
- response vs explanatory plot
- residuals vs explanatory plot
- residuals vs predicted value plot



# Assessing constant variance assumption

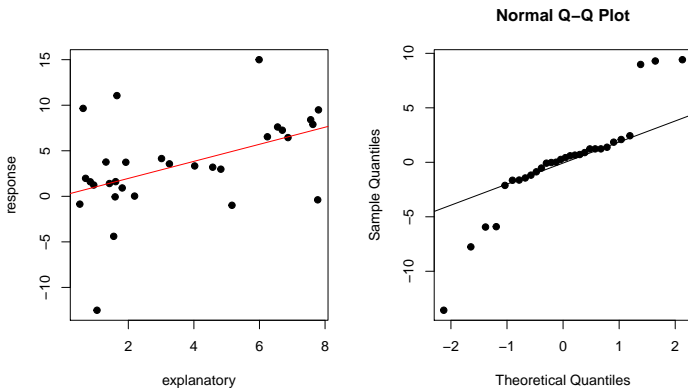
Look for a bugle horn pattern

- residuals vs explanatory plot
- residuals vs predicted value plot



# Assessing normality assumption

Deviations from a straight line in a normal quantile plot (qq-plot)



# Assessing the independence assumption

The main ways that the independence assumption is violated are

- temporal effects
- spatial effects
- clustering effects

Each of these requires a relatively sophisticated plot or analysis and thus, for this course, we will assess the independence assumption using the context of the problem. If one of the above effects are present in the problem, then there **may** be a violation of the independence assumption.

# Influential individuals

In addition to violation of model assumptions, we should be on the lookout for individuals who are influential.

Recall

- if the explanatory variable value is far from the other explanatory variable values, then the individual has high **leverage**, and
- if removing an observation changes the intercept or slope a lot, then the individual has high **influence**.



# Regression analysis procedure

1. Determine hypotheses, i.e. why are you collecting data
2. Collect data (at least two variables per individual)
3. Identify explanatory and response variables
4. Plot the data
5. Run regression
6. Assess regression assumptions
7. Interpret regression output

# Gas mileage

To understand changes in our 2011 Toyota Sienna, we record the miles driven and amount of fuel consumed since our last fill-up. From this we can calculate the miles per gallon (mpg) since our last fill-up.

Understanding changes in mpg through time may give us an indication of problems with our car.

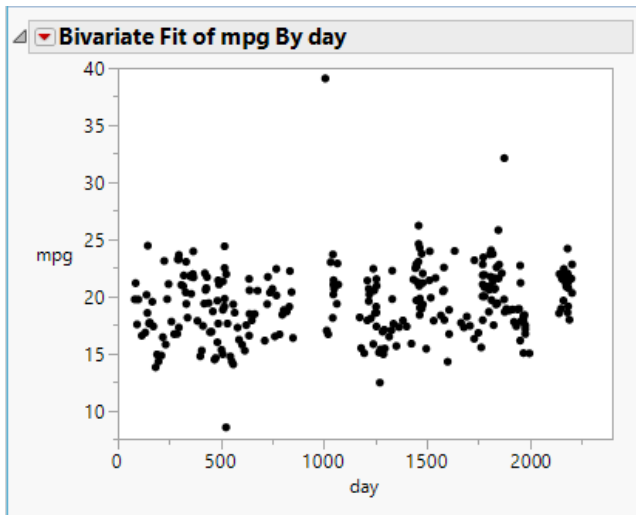
In the following analysis, we use

- miles per gallon as our response variable
- days since purchase as our explanatory variable

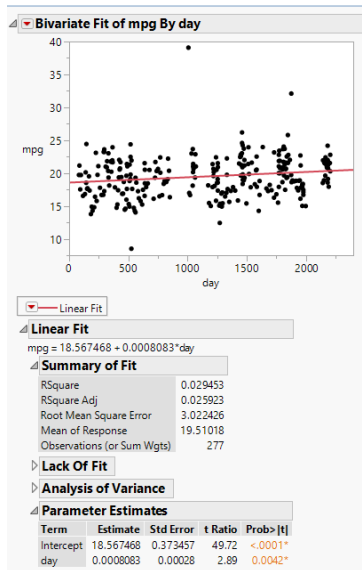
# Example data sheet

date	cost	fuel	mileage	octane	ethanol	notes
6/8	44.52	15.357	284.2	87	10%	Phillips 66
6/10	15.561	45.11	311.6	87	0%	Loaves
6/16	44.20	16.877	311.2	87	10%	Sams
6/21	38.47	14.254	307.6	87	10%	Kum260
6/26	34.00	13.234	284.3	87	10%	Sams
6/29	28.13	10.197	200.1	87	10%	Phillips 66
7/1	33.10	12.451	278.9	87	0%	P. lot
7/2	35.59	13.185	291.0	87	0%	Holiday
7/5	35.46	14.605	326.4	87	0%	Costco
7/11	49.10	17.542	370.9	87	0%	Holiday
7/13	47.40	17.563	366.1	87	10%	Cacey's
7/19	33.90	12.895	239.5	87	10%	Swift Stop
7/19	18.12	6.664	146.6	87	0%	Holiday
7/19	22.10	7.894	190.8	87	0%	Holiday
7/22	27.86	10.322	197.3	87	10%	Genex
7/22	18.24	6.859	145.5	87	10%	Holiday
7/22	18.43	6.778	147.7	87	0%	Holiday
7/23	18.99	7.449	154.3	87	10%	Sams
7/28	24.09	8.762	157.2	87	10%	Phillips 66
8/7	33.23	12.043	257.4	87	10%	Super American
8/10	31.08	11.388	231.0	87	10%	Swift Stop
8/10	17.42	6.455	147.1	87	0	Holiday

# Plot

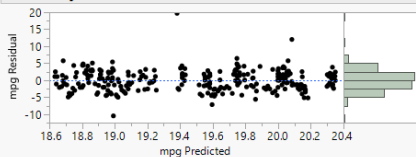


# Regression



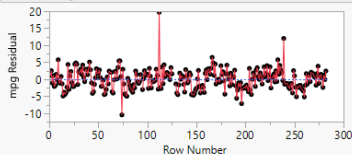
# Residuals

Residual by Predicted Plot

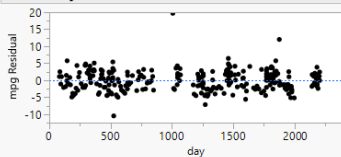


Actual by Predicted Plot

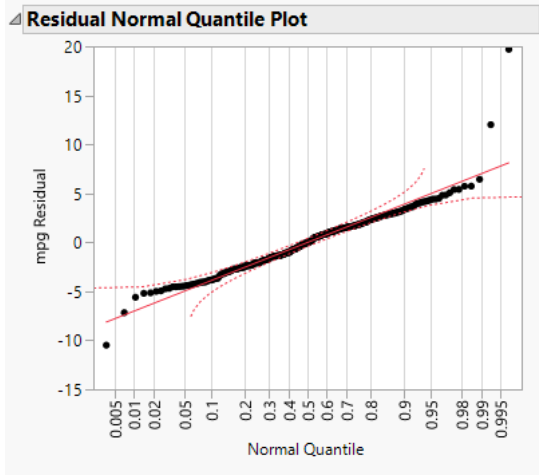
Residual by Row Plot



Residual by X Plot



# Normal quantile plot



# Regression

## Linear Fit

$\text{mpg} = 18.567468 + 0.0008083 \cdot \text{day}$

### Summary of Fit

RSquare	0.029453
RSquare Adj	0.025923
Root Mean Square Error	3.022426
Mean of Response	19.51018
Observations (or Sum Wgts)	277

### Lack Of Fit

### Analysis of Variance

### Parameter Estimates

Term	Estimate	Std Error	t Ratio	Prob> t
Intercept	18.567468	0.373457	49.72	<.0001*
day	0.0008083	0.00028	2.89	0.0042*



# Interpretation

- When the car was purchased (day 0), the predicted miles per gallons was 18.6 mpg.
- Each additional day that passes, the miles per gallons increases by 0.0008 mpg on average. Over the course of a year, this is an increase of 0.29 mpg on average.
- Only 2.9% of the variability in miles per gallon is explained by day.

# Confidence intervals

To construct a  $100(1 - \alpha)\%$  confidence interval, we use the generic formula

$$\text{estimate} \pm t_{n-2, \alpha/2} \text{SE}(\text{estimate})$$

Suppose we are interested in 90% confidence intervals for the intercept and slope. We have

$$t_{275, 0.05} < t_{100, 0.05} = 1.66.$$

Thus, a 90% confidence interval for the intercept is

$$18.567468 \pm 1.66 \times 0.373457 = (17.9, 19.2)$$

and a 90% confidence interval for the slope is

$$0.0008083 \pm 1.66 \times 0.00028 = (0.0003, 0.0013).$$

# Confidence interval interpretation

- Intercept:

- We are 90% confident the true mean miles per gallon on the day of purchase (day 0) was between 17.9 and 19.2 miles per gallon.
- If we repeat this confidence interval construction procedure, 90% of the intervals constructed will contain the true value.
- If we construct 100 intervals, on average 90 of the intervals will contain the true value.

- Slope:

- We are 90% confident the average daily increase in miles per gallon is between 0.0003 and 0.0013 miles per gallon.
- If we repeat this confidence interval construction procedure, 90% of the intervals constructed will contain the true value.
- If we construct 100 intervals, on average 90 of the intervals will contain the true value.

## Bayesian interpretation of credible intervals:

- Intercept: We believe with 90% probability that the true mean miles per gallon on the day of purchase (day 0) was between 17.9 and 19.2 miles per gallon.
- Slope: We believe with 90% probability that the average daily increase in miles per gallon is between 0.0003 and 0.0013 miles per gallon.

# Hypothesis tests

JMP reports two  $p$ -values:

Parameter Estimates				
Term	Estimate	Std Error	t Ratio	Prob> t
Intercept	18.567468	0.373457	49.72	<.0001*
day	0.0008083	0.00028	2.89	0.0042*

These correspond to the hypothesis tests

$$\text{Intercept} \quad H_0 : \beta_0 = 0 \quad \text{vs} \quad H_a : \beta_0 \neq 0$$

$$\text{day} \quad H_0 : \beta_1 = 0 \quad \text{vs} \quad H_a : \beta_1 \neq 0$$

To obtain the one-sided  $p$ -values, you need to divided the  $p$ -value in half and, if the alternative is **not** consistent with the estimate, subtract from 1. So the

Hypotheses		$p$ -value
$H_0 : \beta_0 = 0$	vs $H_a : \beta_0 > 0$	$< 0.0001$
$H_0 : \beta_1 = 0$	vs $H_a : \beta_1 < 0$	$0.9979$

# Hypothesis test decision and conclusion

At significance level  $\alpha = 0.1$ :

- Intercept:  $H_0 : \beta_0 = 0$  vs  $H_a : \beta_0 > 0$ 
  - Decision: Since  $p < 0.0001 < 0.1$ , we reject the null hypothesis.
  - Conclusion: There is statistically significant evidence that the mean miles per gallon on day of purchase (day 0) is greater than 0.
- Slope:  $H_0 : \beta_1 = 0$  vs  $H_a : \beta_1 < 0$ 
  - Decision: Since  $p = 0.9979 > 0.1$ , we fail to reject the null hypothesis.
  - Conclusion: There is insufficient evidence that the average daily change in miles per gallon is less than 0.