Regression Modelling

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Objectives

- Correlation
- Linear Regression
- Multi-Variable Regression
- Interpretation and Application
- NOT focussing on theory

What Is Regression?

... and why do we use it?

What Is Regression?

What Is Regression?

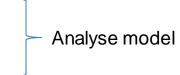
- A statistical analysis that attempts to predict the effects of one or more variables on another variable
- Correlation is a mutual relationship between 2 or more variables

The Process

- Scatter plots
- Measure the degree of linearity between two variables
- Quantify this relationship

What do we assume?

- Assume a causal relationship
- Equation of line of best fit
- Test the significance
- Analyse residuals
- Predication



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46.1538 94.4872 42.8205 91.4103 40.7692 88.3333 38.7179 84.8718 35.641 79.8718 33.0769 77.5641 28.9744 74.4872 26 1538 71 4103 23.0769 66.4103 22.3077 61.7949 22.3077 57.1795 23.3333 52.9481 29.4872 51.0256 32.8205 51.0256 35.3846 51.4103 44,1026 52,9481 46.6667 54.1026 50 55.2564 53.0769 55.641 56.6667 56.0256 59.2308 57.9487 61.2821 62.1795 61 5385 66 4103 61 7949 69 1026 54.8718 49.8718 52.5641 46.0256 49.4872 42.1795 51.0256 44.1026 45.3846 36.4103 42.8205 32.5641 38.7179 31.4103 35.1282 30.2564 32.5641 32.1795 30 36,7949 33.5897 36.6667 38 2051 49 1026 29 7436 36 0256 29 7436 32 1795 30 29.1026 32.0513 26.7949 35.8974 25.2564 41.0356 35.356 44,1026 25,641 47.1795 49.4872 31.4103 53.5897 37.5641 55.1282 40.641 56.6667 42.1795 59.2308 44.4872 62.3077 46.0256 67 9487 47 9481 70 5128 53 718 71 5385 71.5385 64.4872 69.4872 69.4872 46.9231 79.8718 49 2051 94 1026 50 85.2564 53.0769 85.2564 55.3846 86.0256 56.6667 86.0256 56.1538 82.9483 53.8462 80.641 51 2821 78 718 47.9487 77.5641

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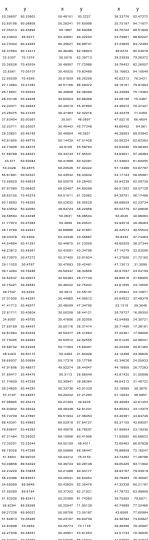
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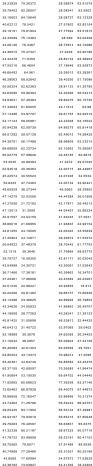
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73.03

57 144

53.711

54.20

52.75

48.563

53.08

56,47538 79,16784

49.57091 79.03812

53.1746t 84.26057

45.272 85.23708

49.96491 83.70832

50.75211 82.90195

x y

51.20381 83.33978

58.97447 85.49982

51.87207 85.82974

48.17992 85.04512

41.6832 84.01794

37.89041 82.56749

39.54897 80.8126

39 64957 82 66453

34 7506 80 01109

x y

47.6952 95.24119

44.60998 93.07584

43.85638 94.08587

41.57892 90.30357

49.17742 96.61052

42.65225 90.56064

30 12331 81 1443

31 9155: 79 79041

27 88771 75 44558

27 54821 75 16476

20.20978 67.5166

24.63081 53.39574

21.03791 49.8293

28.08962 60.51221

31 13571 61 43191

37.70084 68.4184

51,2624 49,66651

43.28671 43.05686

55.90592 54.25267

50.19896 52.22502

55.76151 54.0305

58.32314 60.03106

70 1714/ 70 28222

60 60025 59 75331

63.76027 61.18564

55.2413 52.59357

44,75951 41,97461

51 60061 50 03247

36.46588 35.5241

37 58001 38 83157

36.61972 36.81141

36.16341 33.81931

33.16776 31.24774

42.64498 40.15432

47 22501 44 57565

48 39237 50 33153

27 24515 29 81951

30 40431 29 07954

30.0808 27.58802

33.98061 35.03306

34.25061 34.44159

49.89955 19.56408

55 58987 26 29925

55.48206 22.82747

59.02508 28.32209

62 42377 33 4866

47.84991 48.55662

52.26856 52.80524

77 74685 47 37183

80 35291 47 8003

60 33750 63 03496

61.65872 64.10002

71 54504 72 65773

49.04096 77.36258

#1 99697 PO 11361

57.96898 85.35605

54.85566 83.7834

54.1332 83.794

51.30882 80.13639

61.12035 88.70405

55.56381 85.81038

55.04411 83.54141

50.44411 78.19581

x y

52.87202 97.3432

59.01414 93.5748

56.37511 96.30515

37.8392 94.35944

39.88537 90.63466

44.07741 84.12585

31 49703 67 91256

25 9526 73 03318

38 43472 82 29404

24 99748 75 44158

32.24628 70.11059

21,7961 49,4995

24.15049 47.1885

20.932 51.64624

35.43553 34.96104

44.37738 57.68474

46.38607 51.00216

42,49544 56,90368

44.30052 58.13296

46.98828 55.4206

64.091 68.65093

63.01687 68.22526

67 30258 64 41346

67 22596 62 47641

45.41264 53.8708

54.02061 47.2699

50.7133 48.77221

55.04555 45.98627

37.5437 32.88402

39.7959 28.44322

28.52339 39.79667

31.39965 39.26526

39.24157 33.36142

27.71406 42.2500

35.40474 31.93916

51 64223 50 83834

34 20762 33 59381

36 52349 35 67105

28.4192 21.31575

26.1605 23.0422

40.43533 27.64007

40.12373 30.07316

42 18481 28 57383

40.00266 27.48751

57.12523 42.73902

60 37574 41 35646

55 1059 46 14905

61,73727 38,58484

55.74124 46.65358

76 77561 55 2546

71 88708 59 26699

73 78837 58 62081

75.13968 54.24455

64.59632 65.95251

46.90954 85.60226

40 0051 91 5050

49.80699 82.08503

48.86324 83.15836

66.74819 84.00483

62.35977 90.2436

47.3749 83.47156

47.44647 82.62212

53 38528 78 87114

49.45271 79.9306

x y

65.81554 95.5883

65.67227 91.933

39.00272 92.2618

37.7953 93.5324

35.5139 89.5991

39.21945 83.5434

31.5882 77.0280

32.65778 80.1138

31 5415 77 5108

30 98714 63 9440

27.43963 65.7444

28.8094 59.62

27.93952 50.9850 33.84995 46.2844

32.62846 54.0637

32 43864 49 9030

34.64455 54.4441

35.6754 49.6256

64,99459 49,4621

66.12915 61.6437

67.50178 60.8502

67 62951 61 4291

63.91006 62.9723

88 54938 - 87 7255

65.27839 44.8203

34.99766 28.6329

33.92377 53.5586

66.00445 52.6846

37.02326 32.1184

34.69827 36.4753

34.86344 30.9392

31.09007 32.3837

34.92131 35.4184

30.34336 32.5245

33.23491 38.5024

35 43004 47 7075

33 87351 37 0560

28 58414 34 6264

31.31955 31.6639

30.79617 23.5190

34.23336 22.4959

36.63079 31.4768

34.54862 26.1110

37.51964 29.2676

62.4257 34.0442

65.44553 43.2735

64.32563 41.6159

65.26434 35.7175

66.3887 38.8447

68.54066 48.4385

67 66141 51 0685

68.67552 50.4652

69 08171 55 4740

69.04872 66.3660

67 30822 71 590

38.7149 76.3969

35.54809 89.3049

36.71067 86.0749

64.91266 87.6451

64.93593 87.0721

67.02655 76.2225

65.31403 77.9387

64.29522 84.1258

38.79801 74.6106

| 41.02697 | 76.40959 | 34.7506 | 80.01109 |
|----------|----------|----------|----------|
| 34.57531 | 72.72484 | 27.56084 | 72.84783 |
| 31.1686 | 69.24542 | 24.63554 | 71.61071 |
| 32.64425 | 70.73256 | 20.95946 | 66.0415 |
| 26.66665 | 62.98097 | 20.68915 | 62.72131 |
| 26.45926 | 62.63467 | 19 2882 | 62.06306 |
| 25.53801 | 60.63024 | 20.0245 | 61.34262 |
| 26.32071 | 62.39628 | 35.46952 | 43.11588 |
| 26.16339 | 62.02681 | 36.89431 | 47.70656 |
| | | | |
| 25.21379 | 60.2734 | 39.05555 | 55.54697 |
| 26.85953 | 63.14183 | 46.95708 | 65.24041 |
| 31.76067 | 69.79766 | 37.31045 | 45.25875 |
| 39.79329 | 75.80796 | 40.00967 | 60.98658 |
| 45.61744 | 78.11997 | 48.01435 | 65.71282 |
| 53.77146 | 79.30293 | 53.70378 | 66.38948 |
| 56.45143 | 79.24711 | 63.0675 | 64.035 |
| 66.09372 | 77.01782 | 62.04802 | 63.84586 |
| 56.92584 | 79.20632 | 59.83997 | 64.47676 |
| 58.98825 | 78 91726 | 55.16094 | 85 23731 |
| 57.90227 | 79.08087 | 61.27975 | 65.7664 |
| 64.81287 | 77.48694 | 60.83491 | 64.60377 |
| | | | |
| 60.34979 | 78.72945 | 61.52055 | 64.79186 |
| 48.89056 | 16.80654 | 36.91654 | 41.09525 |
| 47.54951 | 17.0714 | 38.5022 | 41.56716 |
| 30.88943 | 26.32904 | 48.66437 | 30.68067 |
| 28.97284 | 29.05047 | 50.28525 | 30.33792 |
| 35.16136 | 22.78685 | 42.27632 | 34.52764 |
| 45.87037 | 17.48939 | 54.03178 | 29.67235 |
| 32.83142 | 24.79754 | 37.3293f | 39.60204 |
| 39.48557 | 20.03305 | 41.38952 | 37.29606 |
| 33.44618 | 24.19064 | 40.07467 | 34.62369 |
| 21.86358 | 46.05259 | 35.34968 | 47.14107 |
| 25.16587 | 35.76661 | 34.7637 | 47.6248 |
| 25.16587 | 51.28771 | 37.02661 | 44.46229 |
| | | | |
| 25.58606 | 60.84658 | 36.45557 | 40.79184 |
| 26.31431 | 33.29717 | 35.53766 | 48.72939 |
| 33.0373 | 24.56446 | 20.40895 | 32.20303 |
| 36.46889 | 21.87611 | 23.49571 | 25.32247 |
| 26.51355 | 32.91875 | 29.55754 | 21.36478 |
| 36.02709 | 22.19544 | 33.00822 | 15.98507 |
| 42.68104 | 18.54254 | 53.9804 | 29.35099 |
| 45.04108 | 17.73452 | 52.23431 | 29.71167 |
| 47.35242 | 17.13475 | 59.50308 | 30.66967 |
| 53.84619 | 16.38732 | 41.16378 | 34.31576 |
| 55.12964 | 16 35325 | 48 99304 | 32 03036 |
| 52.18812 | 16.41822 | 59.26928 | 29.6407 |
| | | | |
| 58.67266 | 16.66337 | 45.46918 | 33.83119 |
| 82.14431 | 33.19522 | 62.69127 | 30.29037 |
| 75.68657 | 24.74151 | 73.42867 | 48.57786 |
| 85.34496 | 52.88642 | 70.84642 | 52.28225 |
| 83.84465 | 58.61606 | 71.53901 | 45.52181 |
| 85.66476 | 45.54275 | 67.62087 | 38.00656 |
| 77.931 | 68.69006 | 72.47095 | 51.12213 |
| 73.03183 | 73.12057 | 64.81224 | 62.81092 |
| 70.1028 | 75.68321 | 60.85368 | 65.49915 |
| 57.14469 | 79.1941 | 67.7895 | 61.3637 |
| 46.96246 | 78.66656 | 41.60956 | 83.84869 |
| 50.56493 | 79.0773 | 53.00302 | 84.6748 |
| 53.71968 | 79.30694 | 54.71417 | 84.04313 |
| 54.20023 | 79.30594 | 44.29167 | 82,90944 |
| 47.8752 | 79.325 | 49.19172 | 85.87623 |
| | | 49.19172 | 84 54766 |
| 51.47346 | 79.15987 | | |
| 52.75122 | 79.29366 | 51.59985 | 84.81982 |
| 48.56222 | 78.78238 | 54.37972 | 84.24036 |
| 53.08124 | 79.29725 | 46.48073 | 83.51821 |
| 60 47570 | 70 16784 | 62 17464 | 94.20062 |

54.26 X Mean 47.83 Y Mean

16.76 X Standard Deviation 26.93
Y Standard Deviation

-0.06

Correlation Coefficient

Scatter Plots

Plot a scatter diagram and look for evidence of linear trend

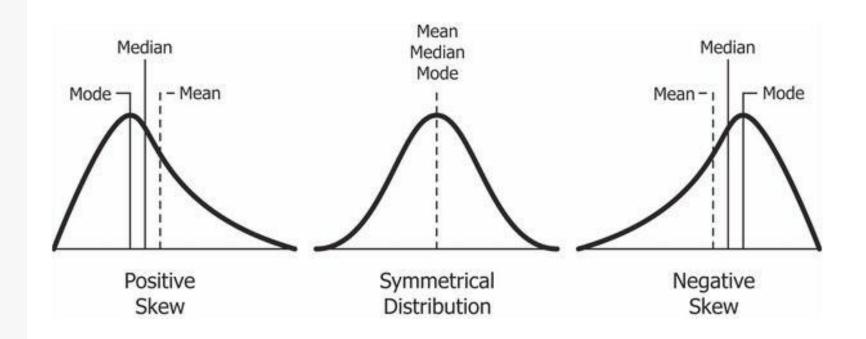
Skew

Skewness, in statistics, is the degree of distortion from the symmetrical bell curve in a probability distribution.

Can be positive, negative or zero, to a varying degree.

Helps to consider the extremes of data, not just the averages, by standardising data.

Skew



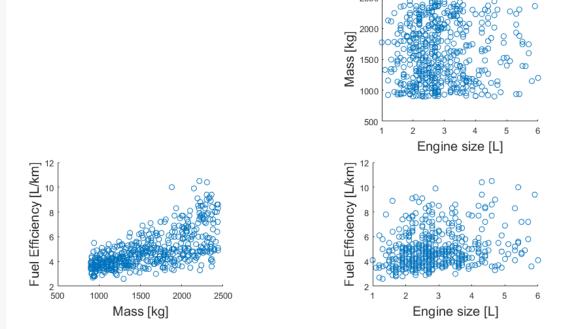
Car Emissions Data Task

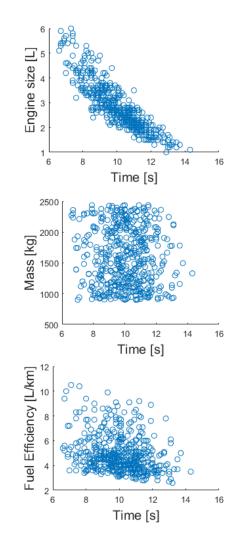
Regression Analysis

Generate and evaluate a model for fuel efficiency in terms of time, vehicle mass, engine size, fuel type and colour

Scatter Plots

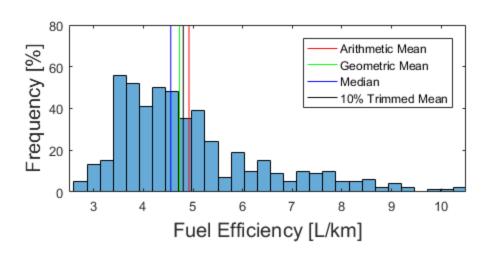
Looking for possible correlation





Histogram

- Looking for skew in the response variable
- Positive skew => mode < median < mean</p>



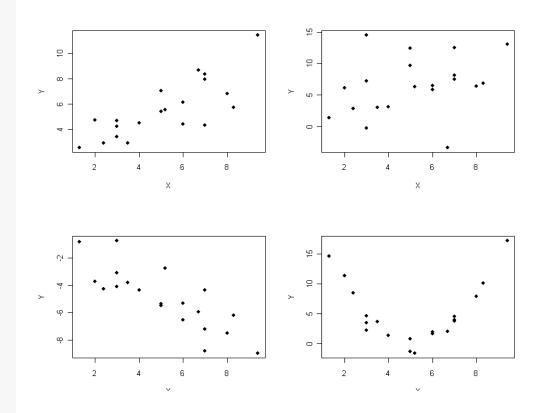
Initial Thoughts...

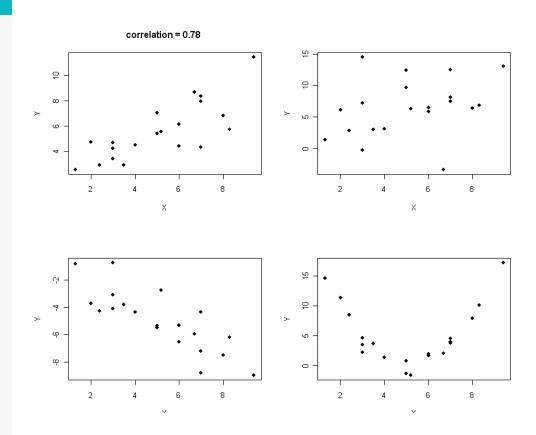
- Correlation exists between fuel efficiency, vehicle mass, acceleration time and engine size and engine size and acceleration time.
- Weaker correlation between fuel efficiency and engine size
- Positive skew, so we should standardise by the geometric mean, $x' = \frac{x \bar{x}_{Geometric}}{\sigma_{Geometric}}$

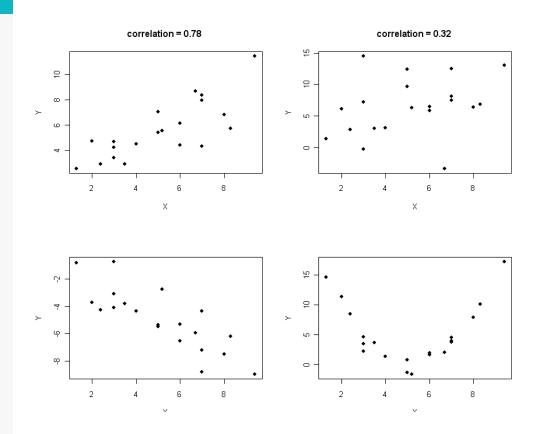
- Having decided that x_i and y_i are paired
- We want to study the relationship between them

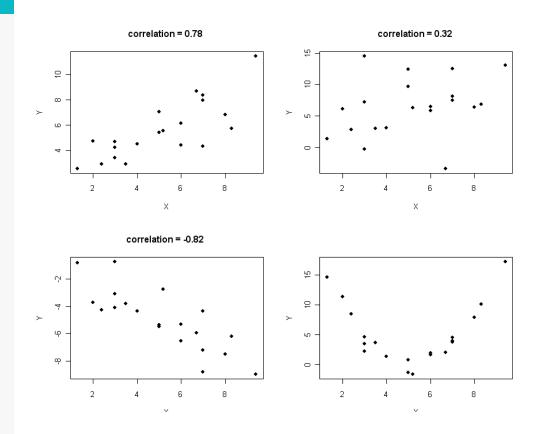
► The Correlation coefficient measures the degree of linear relation between x and y

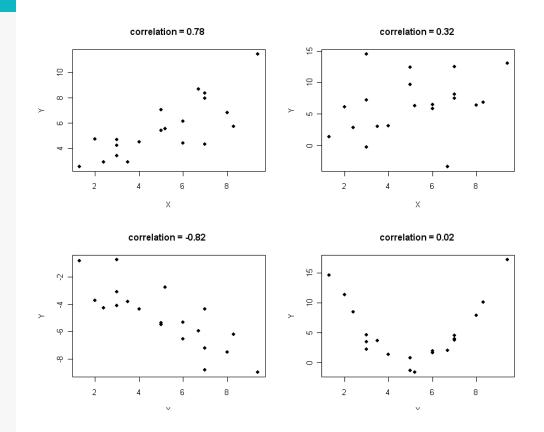
| R | Results | Correlation | Example |
|----|-----------------------|-------------------------|---------------------|
| 0 | Uncorrelated | No correlation | x x x x x x x x x x |
| >0 | Positively correlated | $y = \beta x + \alpha$ | x x x |
| <0 | Negatively corerlated | $y = -\beta x + \alpha$ | x x x x |

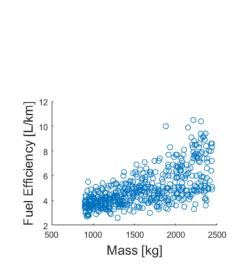


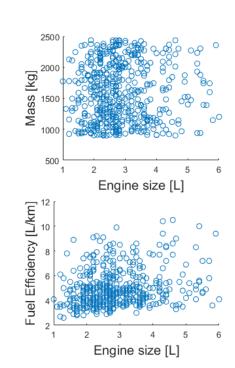


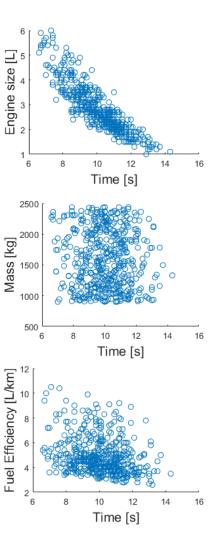


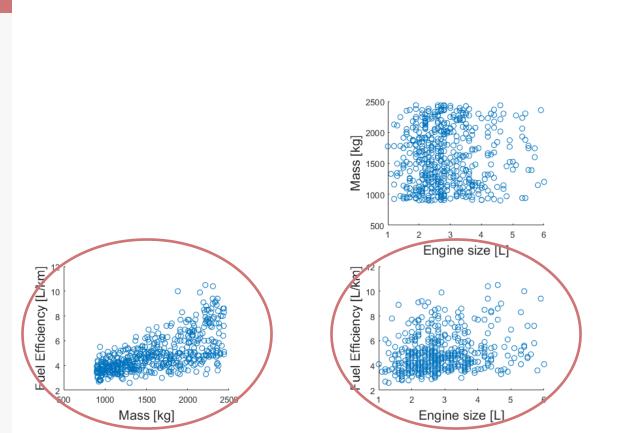


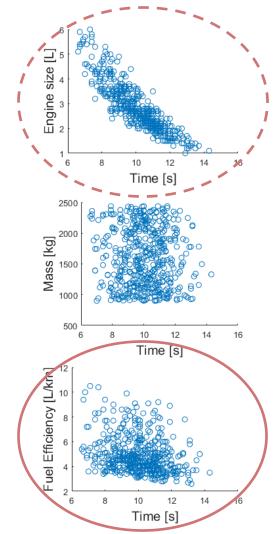








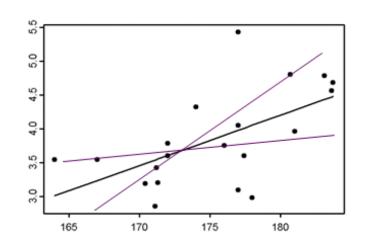




- Not necessarily interested in the correlation coefficient
- Where does weaker correlation exist
- Interaction term may be needed between engine size and time

Regression

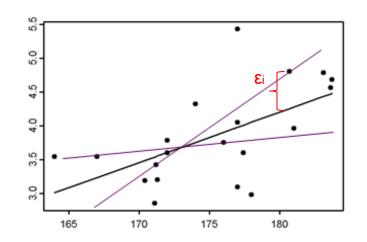
- Let a = estimated α b = estimated β



- a = average fuel consumption at height = 0
- b = increase in fuel consumption for 1cm increase in height

Regression: Residuals

- Our model is not perfect



- Residuals are the error between our model predictions and the actual data
- We assume these are normally distributed

Regression: Least Squares

Find the line of best fit (estimate α and β) to minimise this error

$$\sum_{i=1}^{n} \epsilon_i^2 = \sum_{i=1}^{n} (y_i - \alpha - \beta x_i)^2$$

MATLAB, Python and excel all have libraries to solve this

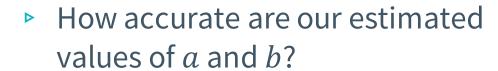
Regression: Least Squares

$$\sum_{i=1}^{n} \epsilon_{i}^{2} = \sum_{i=1}^{n} (y_{i} - \alpha - \beta x_{i})^{2}$$

- ▶ This is an optimisation in terms of α and β
- ightharpoonup The resulting model predicts y values given x

y = a + bx

Regression: Confidence Interval





- 95% CI in b calculated from t scores
- b (b tSE(b), b + tSE(b))



| df/p | 0.40 | 0.25 | 0.10 | 0.05 | 0.025 | 0.01 | 0.005 | 0.0005 |
|------|----------|----------|----------|----------|----------|----------|----------|----------|
| 1 | 0.324920 | 1.000000 | 3.077684 | 6.313752 | 12.70620 | 31.82052 | 63.65674 | 636.6192 |
| 2 | 0.288675 | 0.816497 | 1.885618 | 2.919986 | 4.30265 | 6.96456 | 9.92484 | 31.5991 |
| 3 | 0.276671 | 0.764892 | 1.637744 | 2.353363 | 3.18245 | 4.54070 | 5.84091 | 12.9240 |
| 4 | 0.270722 | 0.740697 | 1.533206 | 2.131847 | 2.77645 | 3.74695 | 4.60409 | 8.6103 |
| 5 | 0.267181 | 0.726687 | 1.475884 | 2.015048 | 2.57058 | 3.36493 | 4.03214 | 6.8688 |
| 6 | 0.264835 | 0.717558 | 1.439756 | 1.943180 | 2.44691 | 3.14267 | 3.70743 | 5.9588 |
| 7 | 0.263167 | 0.711142 | 1.414924 | 1.894579 | 2.36462 | 2.99795 | 3.49948 | 5.4079 |
| 8 | 0.261921 | 0.706387 | 1.396815 | 1.859548 | 2.30600 | 2.89646 | 3.35539 | 5.0413 |
| 9 | 0.260955 | 0.702722 | 1.383029 | 1.833113 | 2.26216 | 2.82144 | 3.24984 | 4.7809 |
| 10 | 0.260185 | 0.699812 | 1.372184 | 1.812461 | 2.22814 | 2.76377 | 3.16927 | 4.5869 |
| 11 | 0.259556 | 0.697445 | 1.363430 | 1.795885 | 2.20099 | 2.71808 | 3.10581 | 4.4370 |
| 12 | 0.259033 | 0.695483 | 1.356217 | 1.782288 | 2.17881 | 2.68100 | 3.05454 | 43178 |
| 13 | 0.258591 | 0.693829 | 1.350171 | 1.770933 | 2.16037 | 2.65031 | 3.01228 | 4.2208 |
| 14 | 0.258213 | 0.692417 | 1.345030 | 1.761310 | 2.14479 | 2.62449 | 2.97684 | 4.1405 |
| 15 | 0.257885 | 0.691197 | 1.340606 | 1.753050 | 2.13145 | 2.60248 | 2.94671 | 4.0728 |
| 16 | 0.257599 | 0.690132 | 1.336757 | 1.745884 | 2.11991 | 2.58349 | 2.92078 | 4.0150 |
| 17 | 0.257347 | 0.689195 | 1.333379 | 1.739607 | 2.10982 | 2.56693 | 2.89823 | 3.9651 |
| 18 | 0.257123 | 0.688364 | 1.330391 | 1.734064 | 2.10092 | 2.55238 | 2.87844 | 3.9216 |
| 19 | 0.256923 | 0.687621 | 1.327728 | 1.729133 | 2.09302 | 2.53948 | 2.86093 | 3.8834 |
| 20 | 0.256743 | 0.686954 | 1.325341 | 1.724718 | 2.08596 | 2.52798 | 2.84534 | 3.8495 |
| 21 | 0.256580 | 0.686352 | 1.323188 | 1.720743 | 2.07961 | 2.51765 | 2.83136 | 3.8193 |
| 22 | 0.256432 | 0.685805 | 1.321237 | 1.717144 | 2.07387 | 2.50832 | 2.81876 | 3.7921 |
| 23 | 0.256297 | 0.685306 | 1.319460 | 1.713872 | 2.06866 | 2.49987 | 2.80734 | 3.7676 |
| 24 | 0.256173 | 0.684850 | 1.317836 | 1.710882 | 2.06390 | 2.49216 | 2.79694 | 3.7454 |
| 25 | 0.256060 | 0.684430 | 1.316345 | 1.708141 | 2.05954 | 2.48511 | 2.78744 | 3.7251 |
| 26 | 0.255955 | 0.684043 | 1.314972 | 1.705618 | 2.05553 | 2.47863 | 2.77871 | 3.7066 |
| 27 | 0.255858 | 0.683685 | 1.313703 | 1.703288 | 2.05183 | 2.47266 | 2.77068 | 3.6896 |
| 28 | 0.255768 | 0.683353 | 1.312527 | 1.701131 | 2.04841 | 2.46714 | 2.76326 | 3.6739 |
| 29 | 0.255684 | 0.683044 | 1.311434 | 1.699127 | 2.04523 | 2.46202 | 2.75639 | 3.6594 |
| 30 | 0.255605 | 0.682756 | 1.310415 | 1.697261 | 2.04227 | 2.45726 | 2.75000 | 3.6460 |
| z | 0.253347 | 0.674490 | 1.281552 | 1.644854 | 1.95996 | 2.32635 | 2.57583 | 3.2905 |
| CI | | | 80% | 90% | 95% | 98% | 99% | 99.9% |

Warning

▶ Do not use your model to predict data outside the range of values in the domain of x

Be cautious of overfitting

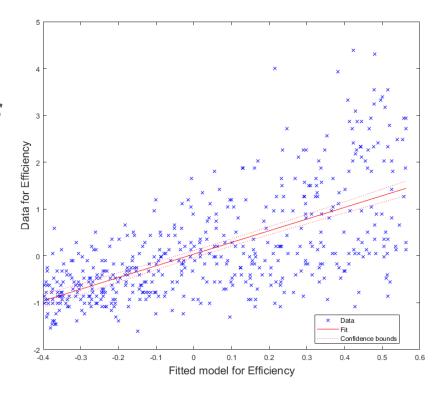
Regression with a Single Predictor Variable

- Model with one predictor:
- $\triangleright \quad Efficiency = \alpha + \beta Mass$

Regression with a Single Predictor Variable

- Model with one predictor:
- $Efficiency = \alpha + \beta Mass$

- Rsquared = 0.4157
- \triangleright MSE = 0.7402
- ▶ AIC = 12705



What do we think of this model?

- Rsquared = 0.4157
- \triangleright MSE = 0.7402
- ▶ AIC = 1271

Multiple Variable Linear Regression?

Multiple Explanatory Variables

- Multiple variables which simultaneously affect output variable
- Interpretation can become increasingly difficult with more variables
- Prevent cofounding and reduce residual variation



Multiple Explanatory Variables

$$y_i = \alpha + \beta_1 x_1, i + \beta_2 x_2, i + \beta_2 x_2, i + \dots + \epsilon_i$$

- \triangleright *i* = number of observations
- $x_{1,i} = i^{th}$ observation of the 1st variable
- $x_{2,i} = i^{th}$ observation of the 2nd variable
- β_1 = the increase in y for a unit increase in x_1

Takes one of a limited number of values

Binary variables take values0 or 1

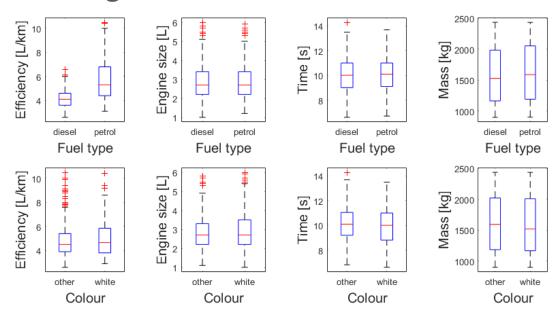
BloodType PoliticalParty

Binary Variables

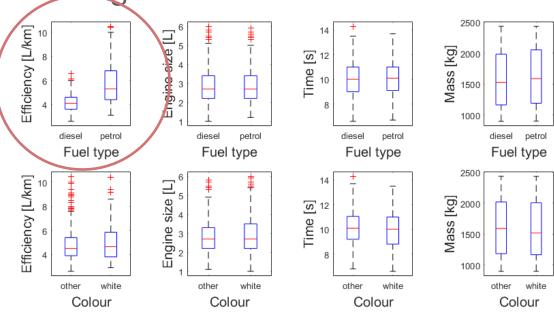


- This model fits the mean of y for each category of x
- $\rightarrow \alpha$ = mean y_i in 1st group
- β = difference in y_i between groups

 Box plots of our data against known categorical variables



Box plots of our data against known categorical variables



Fuel efficiency looks likely to depend on Fuel type to some degree

Fuel type could be a suitable predictor variable in the model

Regression: Least Squares with Multiple Variables

$$\sum_{i=1}^{n} \epsilon_i^2 = \sum_{i=1}^{n} (y_i - \alpha - \beta_1 x_{1,i} - \beta_2 x_{2,i} - \beta_2 x_{2,i})^2$$

- ▶ This is an optimisation in terms of α and $\beta_{1,2,3}$...
- \triangleright The resulting model predicts y values given x

Regression with Multiple Variables

Regression coefficients (b) report the effect of each
 variable while holding all others at their average values

Confounding

- Consider a researcher attempting to assess the effectiveness of drug X, from population data in which drug usage was a patient's choice.
- Data show that gender differences influence a patient's choice of drug as well as their chances of recovery (Y).
- In this scenario, gender z confounds the relationship between X and Y since Z is a cause of both X and Y.

 Gender (Z)





$$y_i = \alpha + \beta_1 x_{1,i} + \beta_2 x_{2,i} + \beta_3 x_{3,i} + \beta_4 (x_{2,i} x_{3,i}) + \dots + \epsilon_i$$

- $\triangleright x_{3,i}$ is a categorical variable
- β_2 = increase in y_i for unit increase in x_2
- $\beta_3 + \beta_4$ = increase in y_i for unit increase in x_2 for an observation in the categorical variable x_3

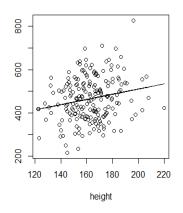
$$\begin{array}{l} {} \blacktriangleright \;\; ArmLength_i = \alpha + \beta_1 Age_i + \beta_2 Height_i + \beta_3 Gender_i \\ + \beta_4 (Height_i Gender_i) + \epsilon_i \end{array}$$

- Gender is the categorical variable
- β_2 = increase in ArmLength for unit increase in Height
- $\beta_2 + \beta_3$ = increase in ArmLength for unit increase in Height for a Female

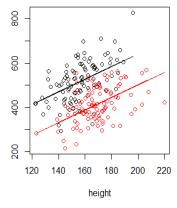
If the interaction term is significant then we should include both the interaction and individual terms

- ► The effect of gender is the different between males and females in this model on the value of height
- This must be found considering the interaction term

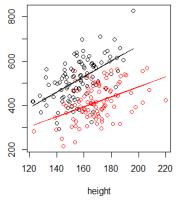
Assuming the effect of height is significant we can have three models depending on whether gender and the interaction are significant.



 $ArmLength_i = \alpha + \beta_1 Age_i + \beta_2 Height_i$



 $ArmLength_i = \alpha + \beta_1 Age_i$ $+ \beta_2 Height_i + \beta_3 Gender_i$



 $ArmLength_i = \alpha + \beta_1 Age_i$ $+ \beta_2 Height_i + \beta_3 Gender_i$ $+ \beta_4 (Height_i Gender_i)$

Improve on the Previous Model by Including more Terms

- Model with one predictor:
- $Efficiency = \alpha + \beta_1 Mass + \beta_2 FuelType$

- Rsquared = 0.6540
- \triangleright MSE = 0.4383
- ► AIC = 1010

Model Checking

Model checking should be performed to avoid erroneous extrapolation of data trends

Residuals

Residuals are the error between the observed and predicted data

Look for trends or patterns in the residuals which indicate an assumption is not valid

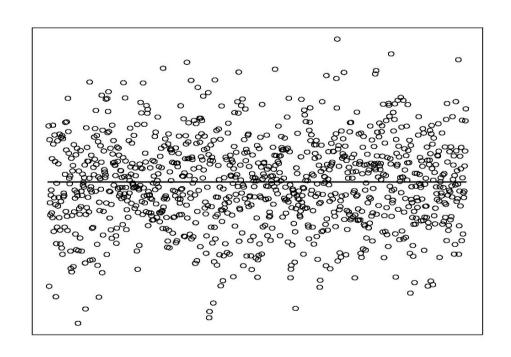
Residual Analysis

- Scatter plots of residuals against fitted values help identify:
 - Non-constant variance
 - Violation of the linearity assumption
 - Potential outliers

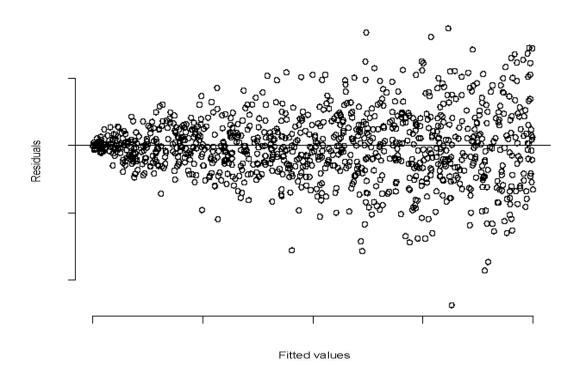
If these assumptions are valid then you will see no trend

Residual Scatter Plot: Satisfactory

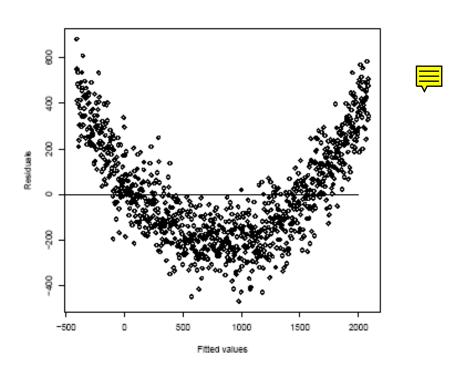
kesiduals



Residual Scatter Plot: Non-constant Variance



Residual Scatter Plot: Non-Linear Relationship



Residual Analysis

- Standardise residuals by dividing by their standard deviation
- They should now be Normal with mean = 0 and variance = 1

Box plots – symmetric?

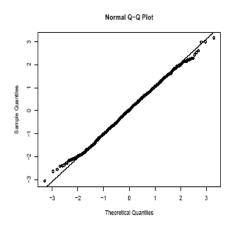
- Proportion of standardised residuals inside percentiles
- Q-Q plot –x = y graph should form

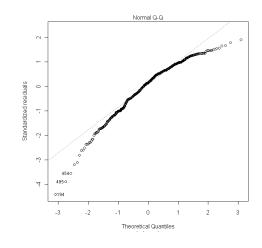
Q-Q Plots and Skew

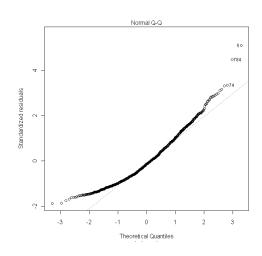
Satisfactory

Negative Skew

Positive Skew







Positive skew => Standardise by the geometric mean

Negative skew=> Standardise

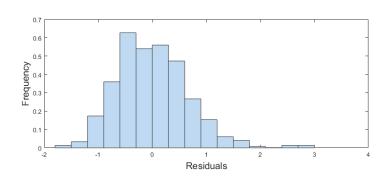
Identifying Outliers

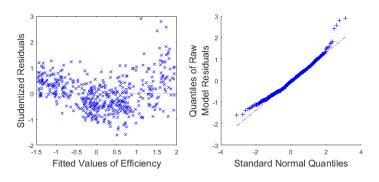
Any points which stand out as having larger residuals than other values should be checked

Cook's Distance is given by most software and measures the influence of each individual point on the model

Residual Analysis

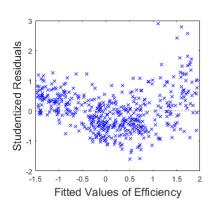
• $Efficiency = \alpha + \beta_1 Mass + \beta_2 FuelType$



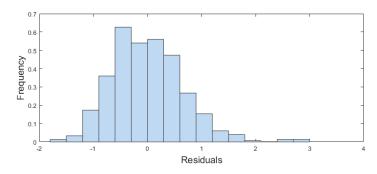


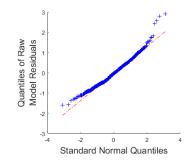
Residual Analysis

Possibly a non-linear relationship?



Maybe very slightly positively skewed?

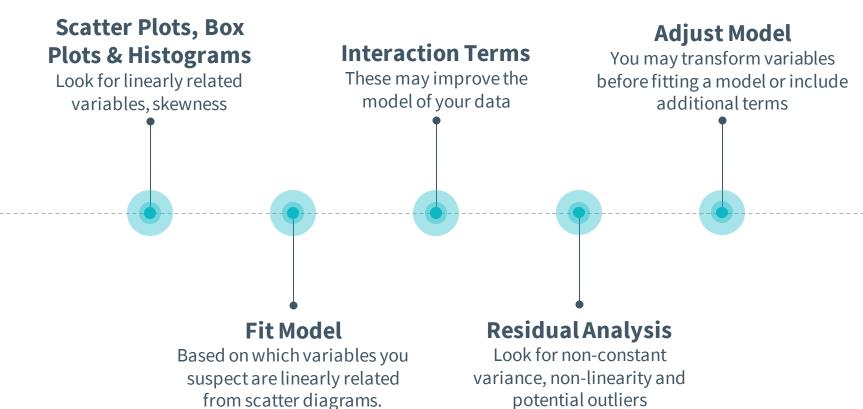




Model Fitting

Model checking should be performed to avoid erroneous extrapolation of data trends

Method



MATLAB Script

Model Fitting

- ▶ Given data on car emissions in terms of vehicle mass, acceleration time, engine size, fuel type and colour.
- Fit models seeking to optimise for:
 - Error
 - AIC relative loss of information

Search for Best Models

- Open file: Regression_Analysis_Car_Emissions
- Use the scatter and box plots to pick a starting model

How to Use the Script

Standardise by the arithmetic or geometric mean

Edit line 115 with your trial model

```
113
         % Insert Model Here
        model = ['Efficiency ~ Mass + FuelType'];
117
        % Other Examples for models
118
         % model = ['Efficiency ~ Mass + EngineSize^2'];
         % AccelTime
119
         % EngineSize
120
121
         % Mass
122
         % FuelType
123
         % Mass:FuelType % Example of an interaction term
124
```

Search for Best Models

- Keep note of the MSE, Rsquared and AIC
- Seek to increase Rsquared
- Seek to reduce MSE and AIC

Things to Try

- Increasing the number of terms
- Including categorical variables
- Using interaction terms (FuelType:AccelerationTime)
- Standardising Method
- Raising terms to a power

How did you do?

Prizes for the 'best' model

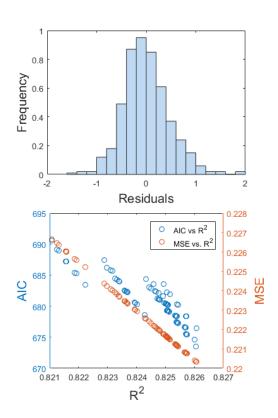
An Optimum Model?

Optimum Model

Running the script:

```
125
126 % Make Model:
127 % [Rmax,mdl,Criteria] = MakeModel(tbl,model);
128 - [CountOpt,Rmax,mdl,Criteria] = MakeOptimumModel(tbl,3)
129
```

- Normally distributed residuals
- AIC and MSE clearly decrease with Rsquared



Model Simplicity

- Model 1
 - Rsquared = 0.8261
 - ► MSE = 0.220
 - ► AIC = 673

```
Efficiency = \alpha + \beta_1 Mass + \beta_2 EngineSize + \beta_3 AccelerationTime + \beta_4 MassFuelType^2
```

- ▶ Model 2
 - Rsquared = 0.8261
 - MSE = 0.220
 - ► AIC = 676

```
Efficiency = \alpha + \beta_1 Mass^3
+ \beta_2 Engine Siz
+ \beta_3 Acceleration Time^2
+ \beta_4 Mass Fuel Type^2
+ \beta_4 Engine Size Fuel Type
```

Further Reading

What are next steps

Next Steps

- K-means cluster classification
- Bootstrapping (Confidence Intervals)
- Introductory courses to Machine Learning (Stanford, coursera)

How did we do?

Please let us know what you thought of this course and how we can improve it

https://forms.gle/HAdcNGkFK5fkCimk6