



## Delivering Data Science In Resources & Energy

### Data Analysis II: Simple Predictions - Regression and statistical model building

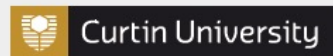
**DAY 5**

**15-Day Data Science Springboard**

**Dr Jeremy Mitchell**  
Data Mettle

**Dr Ying Yap**  
Data Mettle

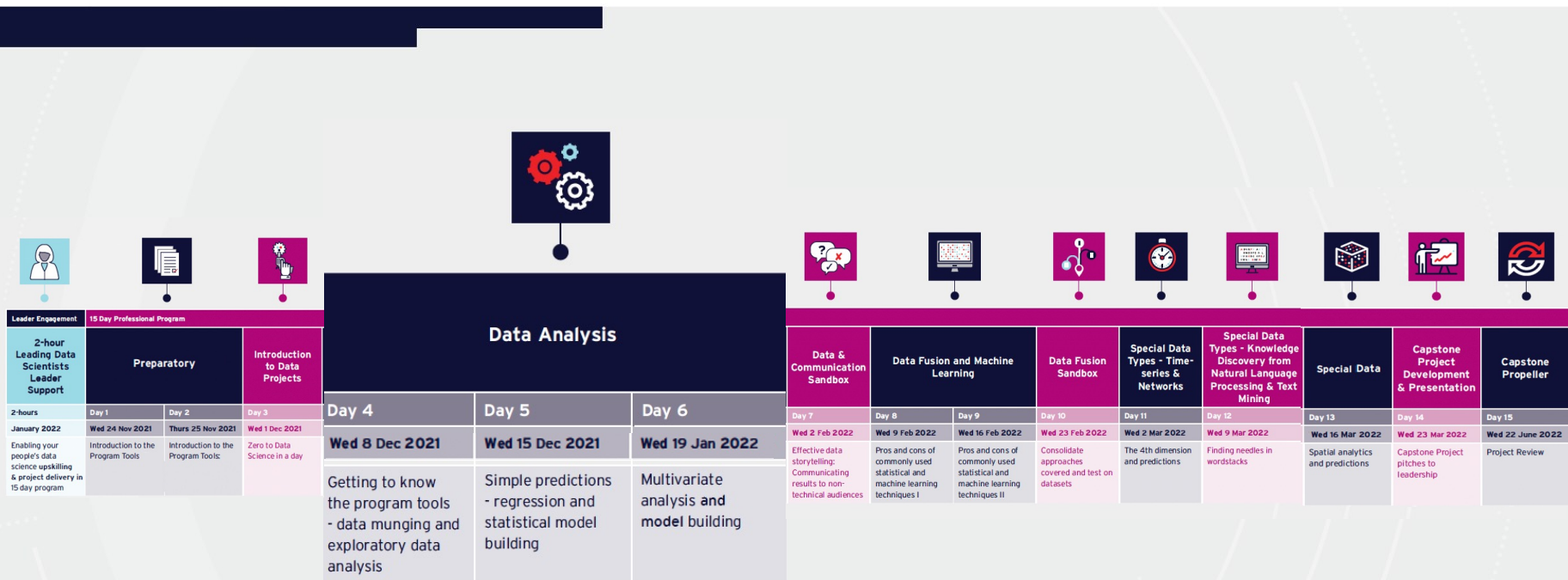
Program partners





# Program Timeline

DAY 4, 5 & 6: Data Analysis





## Q&A, Issues & Announcements



### Before we Get Started

- Resources & Tasks on Github.
- We'll start talking about **projects** a bit this afternoon, and help you start setting up yours
- Make notes about any ideas, perspectives or issues you encounter throughout the day.
- If you have aspects you'd like to go over throughout the day, feel free to post them to the general channel and we'll try to address the straightforward ones as we break.
- We'll come together to discuss before we close out this afternoon.



# Schedule

DAY 5



AWST	AEST	Agenda	Educator
<b>07:30</b>	<b>09:30</b>	<b>Q&amp;A, Issues &amp; Announcements</b>	
07:45	09:45	<a href="#"><u>Models &amp; Regression</u></a>	Jeremy
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17:00	17:00	<b>Close</b>	



# Aims & Learning Outcomes

DAY 5



## Aims

- Perform statistical model building.
- Conduct regression.
- Generate simple predictions - regression.

## Learning Outcomes

- Understand regression as the basis for prediction.
- Understand how outliers and noisy data affect results.
- Understand the impact of missing data and recall practical solutions to work with incomplete data sets.
- Understand how to choose between basic statistical models and evaluate their effectiveness (e.g. linear vs polynomial).
- Have an understanding of hierarchical models as a means of modelling connections between datasets or processes.



# GitHub Content for Today



[github.com / core-skills / 05-simple-predictions](https://github.com/core-skills/05-simple-predictions)

The screenshot shows the GitHub interface for the repository '05-simple-predictions'. At the top, there are buttons for 'Go to file', 'Add file', and a green 'Code' button. Below the 'Code' button, a dropdown menu is open, showing options to 'Clone' (with sub-options for HTTPS, SSH, and GitHub CLI), 'Open with GitHub Desktop', and 'Download ZIP' (which is highlighted with a red box). The repository details show it is on the 'develop' branch, 2 branches ahead, and 1 tag behind master. A pull request by 'morganjwilliams' is also visible. The file list includes 'data', 'notebooks', and 'program'.



# GitHub – Program Notes



[github.com / core-skills / 05-simple-predictions / program / 00\\_overview.md](https://github.com/core-skills/05-simple-predictions/program/00_overview.md)

## Overview

[Overview](#) | [Data Culture](#) | [From Here to There](#) | [Data Projects](#) | [Data Exploration](#) | [Closeout](#)

## Aim

Provide an overview of a 'typical' data science workflow.

## Learning Outcomes

1. To appreciate what data science is
2. To appreciate the fields data science spans
3. Understand the stages of a data science project and define a mental model of it
4. Analyse the opportunity and potential value of data science in your organisation

## Schedule

AWST	AEST	Agenda
07:30 - 07:45	09:30 - 09:45	Q&A, Issues & Announcements
07:45 - 09:15	09:45 - 11:15	<a href="#">Creating a Data Culture</a>
09:15 - 09:30	11:15 - 11:30	<i>Morning Tea</i>
09:30 - 11:00	11:30 - 13:00	<a href="#">Getting From Here to There</a>
11:00 - 11:15	13:00 - 13:15	



- Open an Anaconda Prompt
- Navigate to where you have the unzipped repository material

```
conda env create -f environment.yml # make new env  
conda activate core05 # activate this env (Windows)  
# make this available to Jupyter as a "kernel"  
python -m ipykernel install --user --name=core05  
jupyter lab # launch Jupyter lab
```





# Binder Backup

master 2 branches 2 tags

Go to file Add file + Code +

morganjwilliams mergerbranch-release/1.1 2 minutes ago 11 commits

File	Description	Time
data	Added notebook for data exploration overview	3 months ago
notebooks	Update file & unified notebook for first exercise	21 minutes ago
program	Updated environment file with minimal package versions and get notebook...	1 hour ago
.gitignore	Update Function Notebooks, .gitignore for processed data	3 months ago
LICENSE	Update License Year	3 months ago
README.md	Point Binder Link to Jupyter	3 hours ago
environment.yml	Updated environment file with minimal package versions and get notebook...	1 hour ago

README.md

## CORE Skills Program - Day 3 - Zero to Data Science

launch binder



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# Models & Regression

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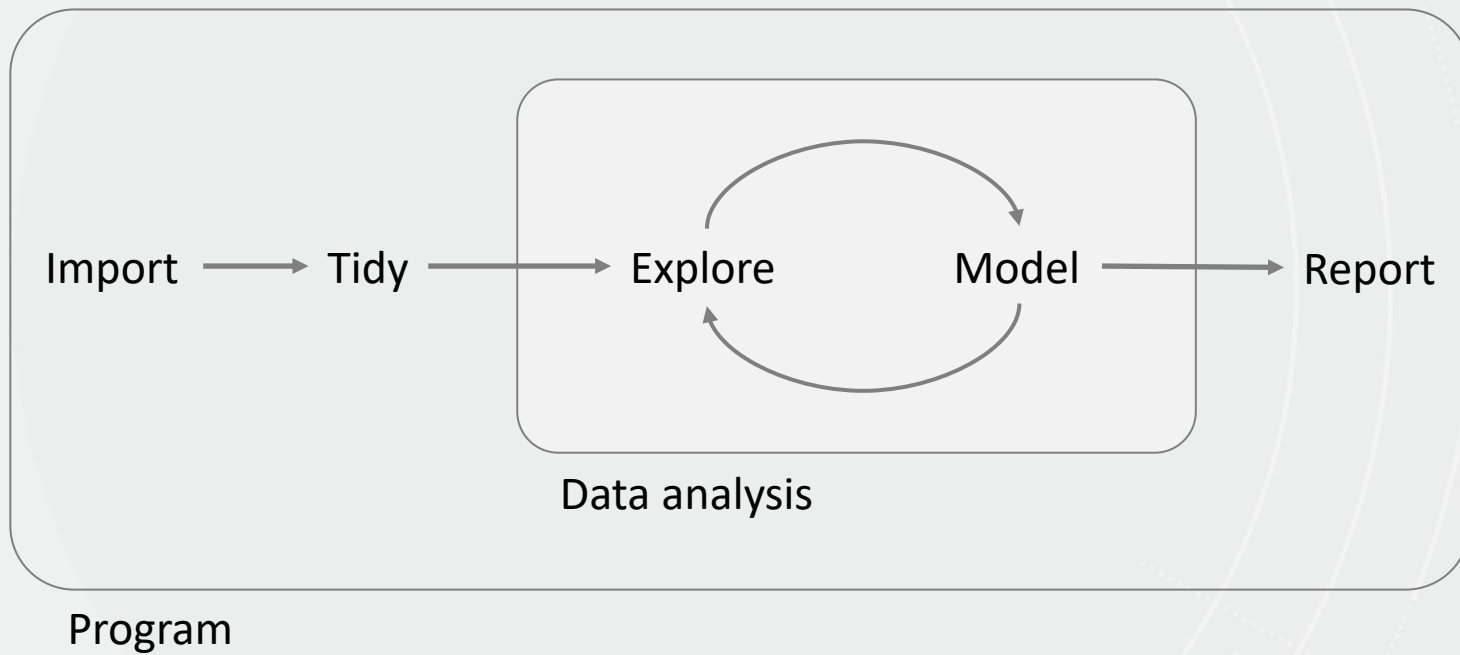


# What are the main steps of the data analysis workflow?

5 min



# Short introduction to data analytics and machine learning





## Exercise: Explore a dataset using pandas and seaborn



**Open `am1-models-and-regression.ipynb`**  
and go through exercise 1



# Categories of machine learning

[scikit-learn.org/stable/tutorial/machine\\_learning\\_map/index.html](https://scikit-learn.org/stable/tutorial/machine_learning_map/index.html)



# Scikit-learn basics and API

- Dataset in tidy format
- Basic API:
  - `model = LinearRegression()`
  - `model.fit(x_train, y_train)`
  - `y_pred = model.predict(x_pred)`





# Linear Regression Model

- The linear regression model is one of the most basic statistical models used in predictive analysis
- The model proceeds by fitting a linear equation to observed paired data to attempt to model the relationship between the two variables
- One variable is commonly referred as to the explanatory variable while the other is considered the dependent variable



# Linear Regression Model

- The equation of a line is

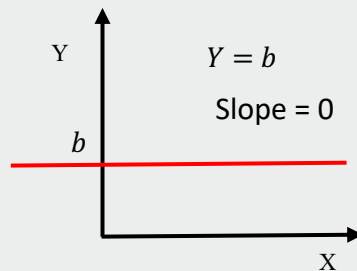
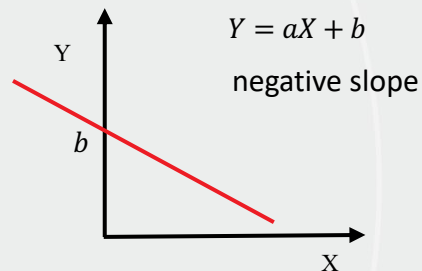
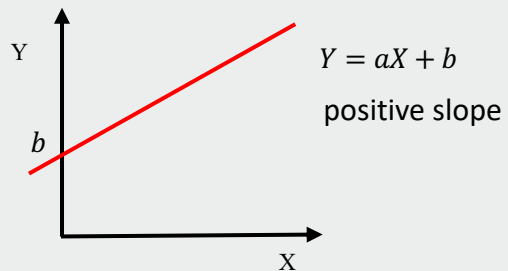
$$Y = aX + b$$

Dependent variable  $\nearrow$   $\nwarrow$  Explanatory variable

- $a$  corresponds to the slope of the line and  $b$  to the intercept
- Parameters  $a$  and  $b$  need to be determined
- Different mathematical approaches can be used to determine the slope and the intercept. This leads to different types of regression



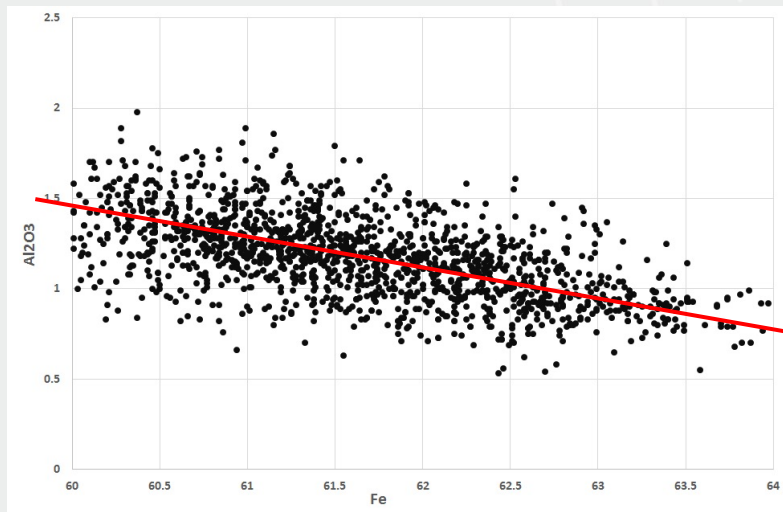
# Linear Regression Model





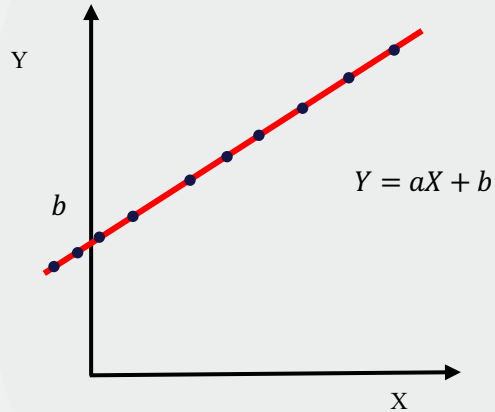
# Linear Regression Model

- Why do we want to fit a line?
- The scatterplot shows the paired data is scattered around a trend line
- The line is indicative of the average value of the dependent variable given the explanatory variable

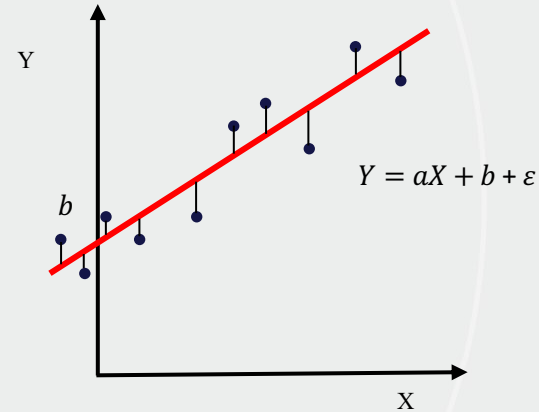




# Linear Regression Model



Values of  $X$  and  $Y$  are perfectly related

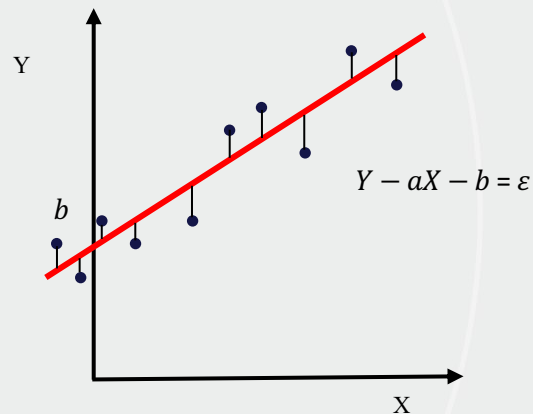


Values of  $X$  are known and it is assumed that  $Y = aX + b$  plus a random term  $\varepsilon$



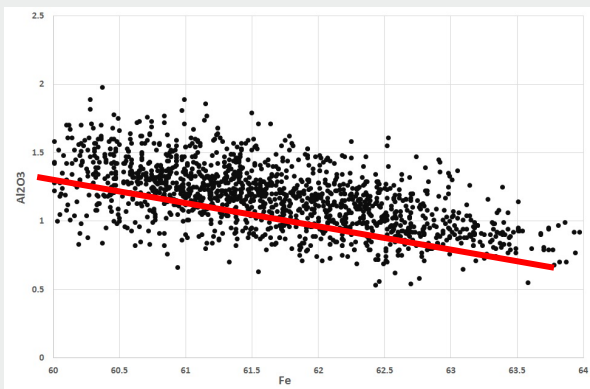
# Statistical assumptions

- A number of assumptions are made on the random error term  $\varepsilon$
- First assumption is that on average the error is equal to zero
- This ensures the model has no bias

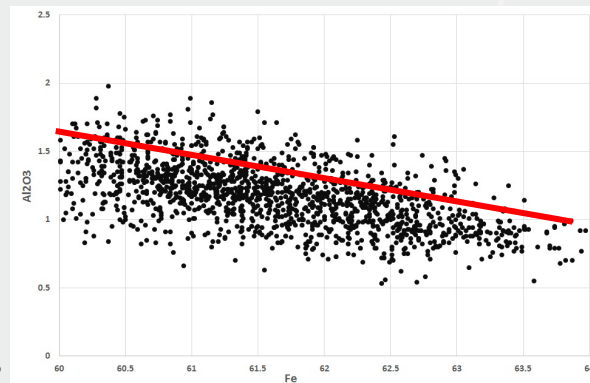




# Statistical assumptions



Positive bias

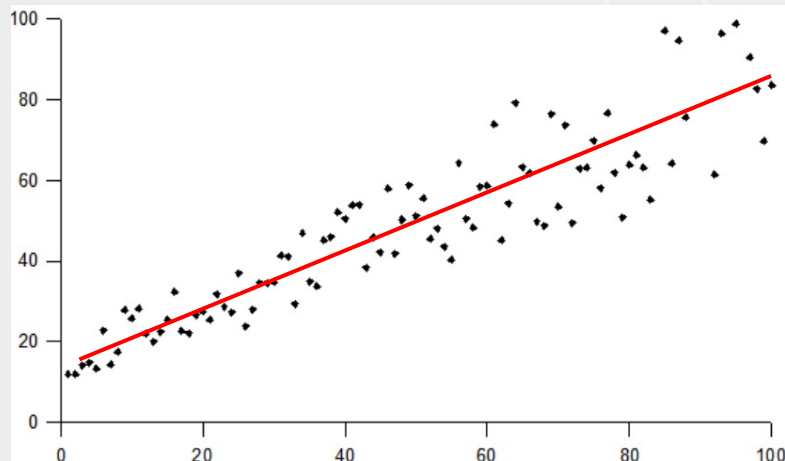


Negative bias



# Statistical assumptions

- Second assumption is that the variance of the random error term is constant
- This is known in statistics as homoscedasticity assumption
- Heteroscedasticity refers to the violation of this assumption







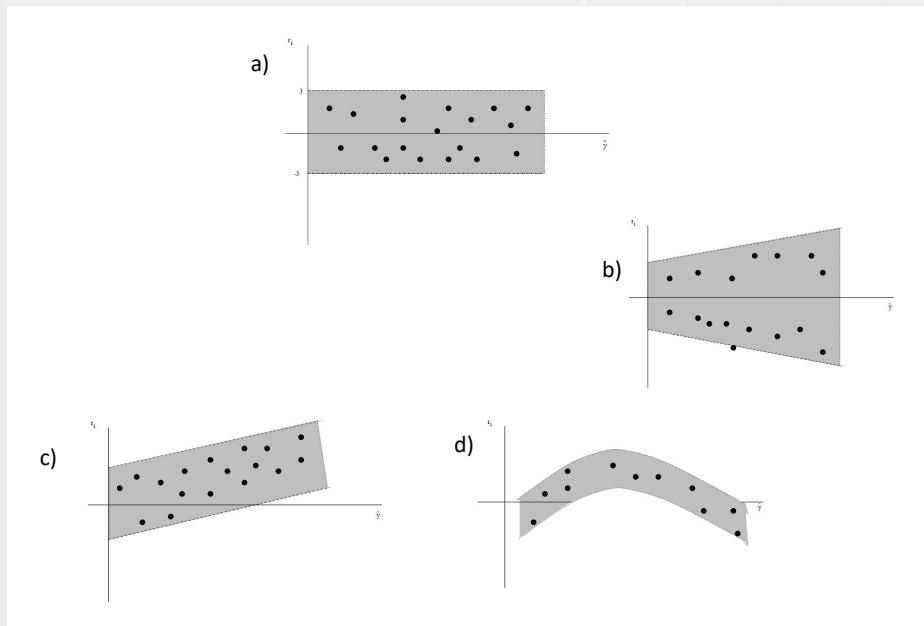
## Statistical assumptions

- Third assumption is that random error term is independent of both the explanatory and dependent variable
- Extremely important assumption and provides a way to assess the goodness of the linear model
- Scatterplot of the error with either the explanatory or the dependent variable should not show any clear pattern



# Statistical assumptions

- a) Appropriate
- b) homoscedasticity violated
- c) and d) are indicative that the linear regression model is not adequate





## Statistical assumptions

- Fourth assumption corresponds to the statistical distribution of the error component
- It is assumed that the distribution of the random error is Gaussian with mean equal to zero and variance  $\sigma^2$

$$\varepsilon \sim N(0, \sigma^2)$$

- The Gaussian assumption allows to derive all the theoretical properties of the linear regression model



## Statistical assumptions

Some of the theoretical properties are:

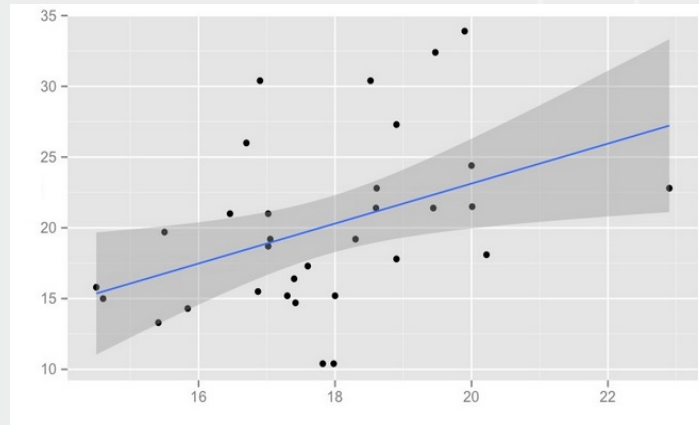
- The dependent variable has Gaussian distribution  
This allows to construct confidence intervals for the estimated values
- The slope  $\alpha$  has Gaussian distribution  
This allows to use statistical tests to assess the “importance” of the explanatory variable

*Statistical tests on the parameters are more important in multivariate problems*



# Statistical assumptions

- Example of a linear regression model with confidence intervals
- Can you anticipate the behaviour of the residuals?
- Would you say the linear model is reasonable?





## Exercise: Linear regression with scikit-learn



Return to **am1-models-and-regression.ipynb**

and go through exercise 2



## Exercise: Perturbing perfect linear data



Return to **am1-models-and-regression.ipynb**

and go through exercise 3



## A look at $R^2$

- $R^2$  is also known as coefficient of determination
- $R^2$  is used as a goodness of fit measure for linear regression models, i.e. to determine how well the regression model fits the data

$$R^2 = 1 - \frac{\sum \varepsilon_i^2}{\sum (y_i - \bar{y})^2}$$

- Proportion of the variance in the dependent variable that the explanatory variable explains





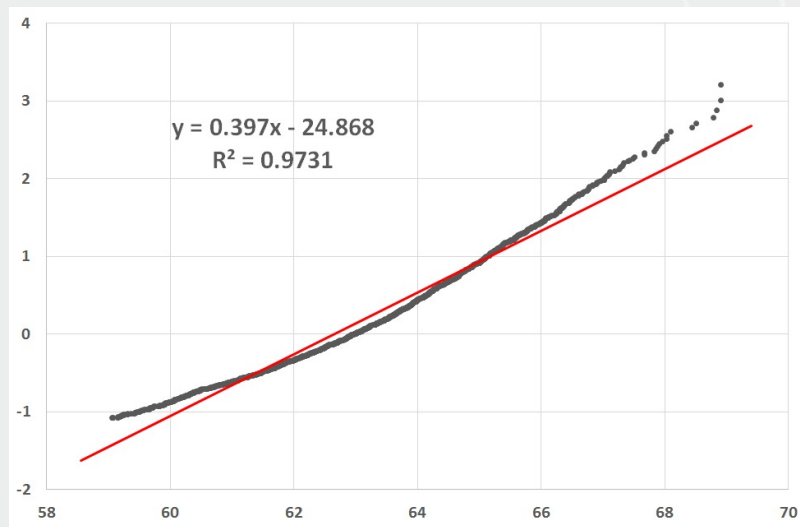
## A look at $R^2$

- $R^2$  has limitations and therefore should not be the only criteria used to assess the goodness of the linear regression model
- Some of its limitations are:
  - Does not account for the number of paired data used
  - Does not indicate if the explanatory variable used is appropriate
  - Does not indicate if the regression used is appropriate
  - Does not indicate if the model is biased



## A look at $R^2$

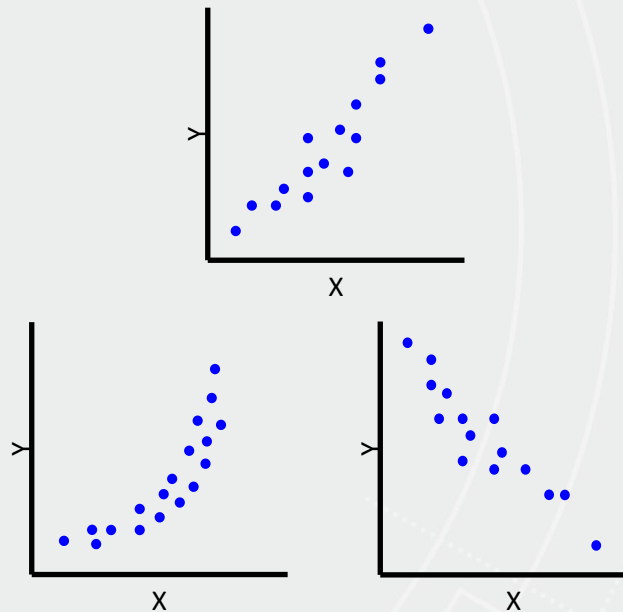
A biased model can have a high  $R^2$  value!





# Correlation and Independence

- Traditional correlation is a measure of the linear relationship between two variables
- A correlation value equal to zero does not mean the variables are not related





# Correlation and Independence

- The concept of statistical independence is given by the factorisation of the joint probability distribution as the product of the marginal distributions
- This means that knowing the value of one of the variables does not tell anything about the value of the other variable
- Independence implies no correlation but the reverse is not true



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**Apply to your  
own Problem**



## If you don't have a dataset

- UCI Machine Learning Repository:  
[archive.ics.uci.edu/ml/index.php](https://archive.ics.uci.edu/ml/index.php)
- Kaggle datasets:  
[www.kaggle.com/datasets](https://www.kaggle.com/datasets)



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# Takeaways & Closeout



## High-level Takeaways From Today

- Understand regression as the basis for prediction.
- Understand how outliers and noisy data affect results.
- Understand the impact of missing data and recall practical solutions to work with incomplete data sets.
- Understand how to choose between basic statistical models and evaluate their effectiveness (e.g. linear vs polynomial).
- Have an understanding of hierarchical models as a means of modelling connections between datasets or processes.



## High-level Takeaways From Today



Your thoughts?





# Capstone Projects



# Capstone Projects



## Update

- How you are shaping up your Project

## Action

- Milestone #1 → L0 in Momentum
- If not done already, update your Working Project Title & Short Project Statement (Problem/Solution/Plan) [here](#).
- Update your Leader




# Daily Feedback

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mentimeter.com

 **Mentimeter**

Please enter the code

1234 5678

Submit

The code is found on the screen in front of you

- What was one thing you like about today?
- What would you like to see more of?



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