Welcome to the CoGrammar Machine Learning Introduction

The session will start shortly...

Questions? Drop them in the chat. We'll have dedicated moderators answering questions.



Data Science Session Housekeeping

- The use of disrespectful language is prohibited in the questions, this is a supportive, learning environment for all - please engage accordingly.
 (Fundamental British Values: Mutual Respect and Tolerance)
- No question is daft or silly ask them!
- There are Q&A sessions midway and at the end of the session, should you
 wish to ask any follow-up questions. Moderators are going to be
 answering questions as the session progresses as well.
- If you have any questions outside of this lecture, or that are not answered during this lecture, please do submit these for upcoming Academic Sessions. You can submit these questions here: <u>Questions</u>



Data Science Session Housekeeping cont.

- For all non-academic questions, please submit a query:
 www.hyperiondev.com/support
- Report a safeguarding incident:
 www.hyperiondev.com/safeguardreporting
- We would love your feedback on lectures: Feedback on Lectures

Skills Bootcamp 8-Week Progression Overview

Fulfil 4 Criteria to Graduation

Criterion 1: Initial Requirements

Timeframe: First 2 Weeks
Guided Learning Hours (GLH):
Minimum of 15 hours
Task Completion: First four tasks

Due Date: 24 March 2024

Criterion 2: Mid-CourseProgress

60 Guided Learning Hours

Data Science - **13 tasks** Software Engineering - **13 tasks** Web Development - **13 tasks**

Due Date: 28 April 2024



Skills Bootcamp Progression Overview

Criterion 3: Course Progress

Completion: All mandatory tasks, including Build Your Brand and resubmissions by study period end Interview Invitation: Within 4 weeks post-course Guided Learning Hours: Minimum of 112 hours by support end date (10.5 hours average, each week)

Criterion 4: Demonstrating Employability

Final Job or Apprenticeship
Outcome: Document within 12
weeks post-graduation
Relevance: Progression to
employment or related
opportunity





Learning objectives

Understand and implement simple linear regression models using Python and scikit-learn.

- Define simple linear regression and its purpose
- Interpret the mathematical equation and assumptions of simple linear regression
- Implement and evaluate simple linear regression models using Python



Introduction to Machine Learning

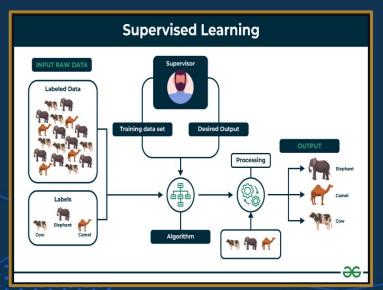
- Machine learning is a way of teaching computers to learn and improve from experience without being explicitly programmed.
- It allows computers to automatically learn and adapt based on data.





Types of machine learning

Supervised learning: The computer learns from labelled data, where both input and output data are provided.

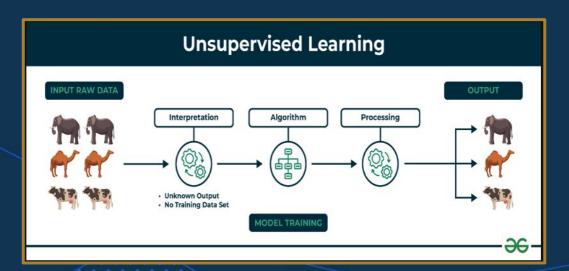


Source: <u>geeksforgeeks</u>



Types of machine learning

Unsupervised learning: The computer learns from unlabeled data, discovering hidden patterns or structures on its own.



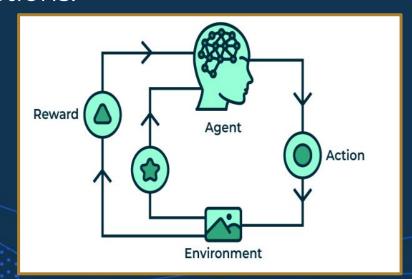
Source: geeksforgeeks





Types of machine learning

Reinforcement learning: The computer learns through interaction with an environment, receiving rewards or penalties for its actions.



Source: geeksforgeeks



Some Applications of machine learning

- Spam email filtering: Identifying and separating spam emails from regular emails.
- Image recognition: Recognizing objects, faces, or scenes in images.
- Recommender systems: Suggesting products, movies, or songs based on user preferences.



Supervised Learning

- In supervised learning, the algorithm learns from labelled data, which consists of input-output pairs.
- The goal is to learn a function that maps input data to the correct output labels.





Types of Supervised Learning

- Regression: Predicting continuous numerical values, such as house prices or stock prices.
- Classification: Predicting discrete categories or classes, such as whether an email is spam or not.



Supervised Learning Algorithms

- Linear regression: Fitting a straight line to data points to make predictions.
- **Logistic regression:** Predicting binary outcomes, such as yes/no or true/false.
- **Decision trees:** Making decisions based on a series of questions or conditions.
- Support vector machines (SVM): Finding the best boundary to separate different classes.
- Neural networks: Mimicking the structure and function of the human brain to learn complex patterns.

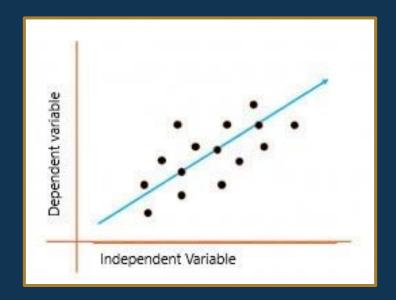


Simple Linear Regression

- Simple linear regression is a method to study the relationship between two variables: an independent variable (x) and a dependent variable (y).
- It helps us understand how changes in the independent variable affect the dependent variable.



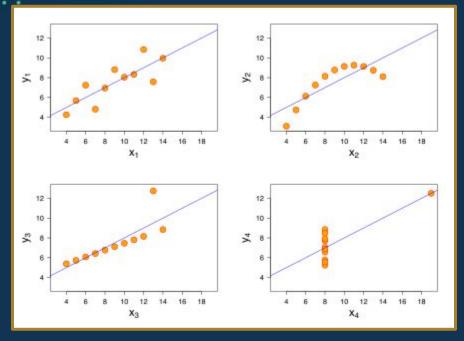




Source: <u>Analytics Vidhya</u>







Source: Wikipedia



Purpose of Simple Linear Regression

- To find the best-fitting straight line that describes the relationship between x and y.
- This line can be used to make predictions about the dependent variable based on new values of the independent variable.





Applications of Simple Linear Regression

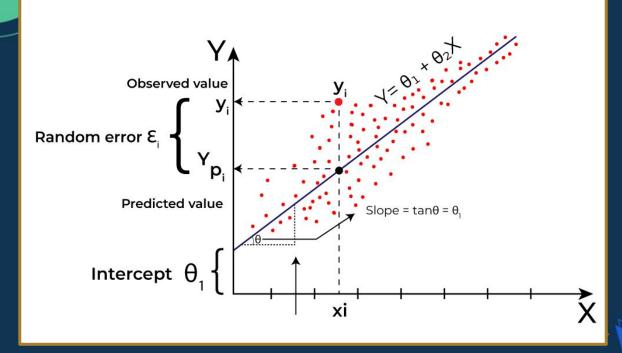
- Sales forecasting: Predicting future sales based on historical data.
- Price prediction: Estimating the price of a product based on its features.
- * Trend analysis: Identifying trends or patterns in data over time.



Math behind Simple Linear Regression

- * The equation is written as: $y = β_0 + β_1x + ε$
 - \triangleright β_0 is the intercept, representing the value of y when x is zero.
 - > β_1 is the slope, indicating how much y changes for a one-unit increase in x.
 - \succ ϵ is the error term, accounting for the variability in y that cannot be explained by x.

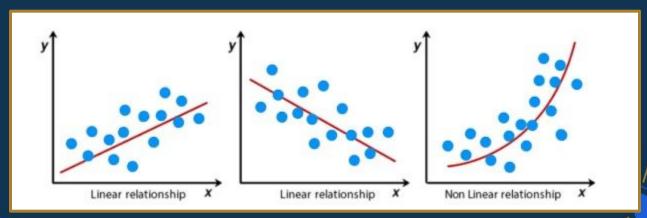




Source: <u>geeksforgeeks</u>

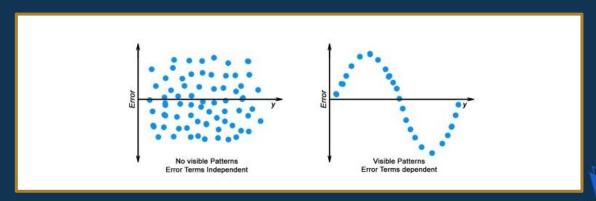


Linearity: The relationship between x and y should be linear.



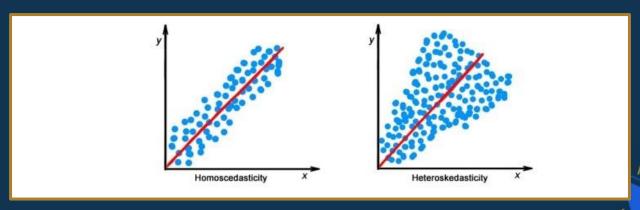


Independence: The observations should be independent of each other.



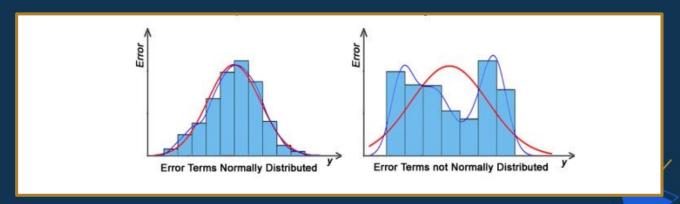


Homoscedasticity: The variability of y should be constant across all values of x.





Normality: The errors should be normally distributed.





Implementing Simple Linear Regression





Scikit-learn

- Scikit-learn is a popular Python library for machine learning.
- It provides simple and efficient tools for data analysis and modelling.

```
from sklearn.datasets import load_diabetes
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
```

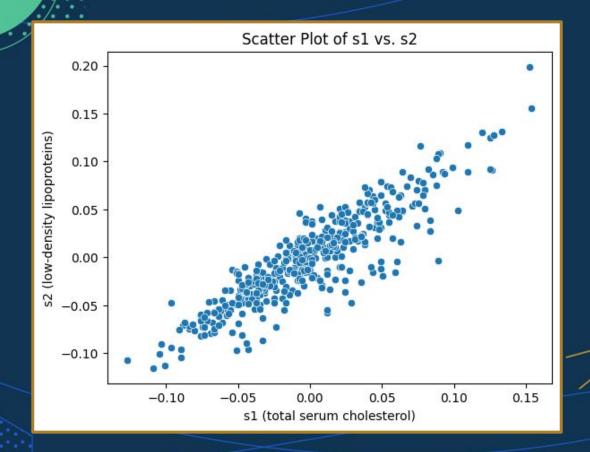


Loading the Diabetes Dataset

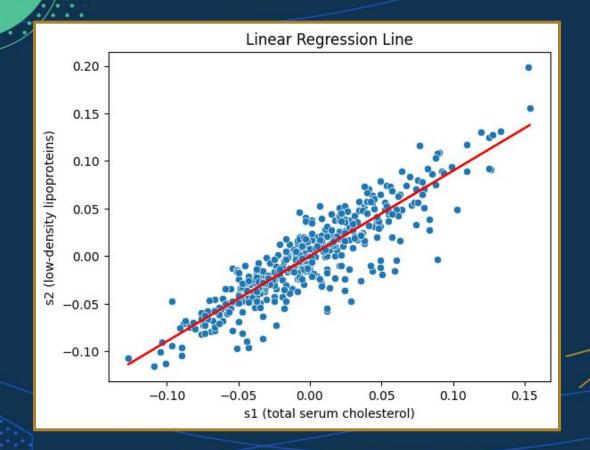
- We'll use the built-in diabetes dataset from scikit-learn for our example.
- The dataset contains information about various medical predictors and a quantitative measure of disease progression.

```
df = load_diabetes(as_frame=True).data
```











Evaluation Metrics:

Mean Squared Error (MSE): 0.00044342882373426217

R-squared (R2) Score: 0.8040044599094562





:Interpretation of Results

- Scatter plot:
 - > The scatter plot visualises the relationship between s1 (total serum cholesterol) and s2 (low-density lipoproteins).
 - It helps assess the linearity and spread of the data points.
- Linear regression line:
 - The red line represents the best-fit line obtained from the linear regression model.
 - ➤ It shows the predicted relationship between s1 and s2 based on the trained model.



Evaluation Metrics

- Mean Squared Error (MSE):
 - MSE measures the average squared difference between the predicted and actual values.
 - A lower MSE indicates better model performance.
- R-squared (R²) score:
 - ➤ R² represents the proportion of variance in the target variable that can be explained by the model.
 - An R² value closer to 1 indicates a better fit of the model to the data.



Evaluation Metrics

$$R^2 = 1 - SS_{error}/SS_{total}$$

Where:

$$>$$
 SS_{error} = $\Sigma (y_i - \hat{y}_i)^2$

$$>$$
 SS_{total} = $\Sigma(y_i - E[y_i])^2$



Evaluation Metrics

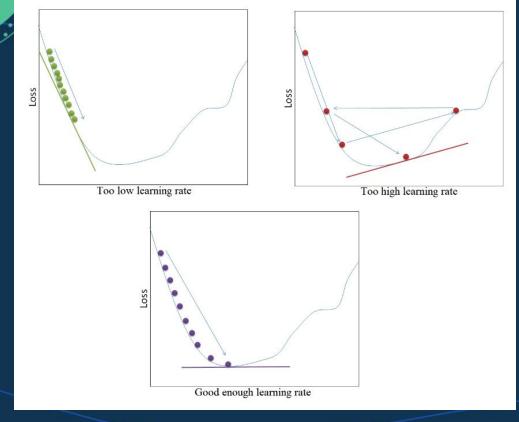
- Accuracy is another commonly used metric for evaluating the performance of a machine learning model, particularly in classification problems.
 - Accuracy = (Number of correct predictions) / (Total number of predictions) * 100%
- While accuracy is more suitable for classification tasks, metrics like Mean Squared Error (MSE) and R-squared (R^2) are used for regression problems.



Parameter Tuning

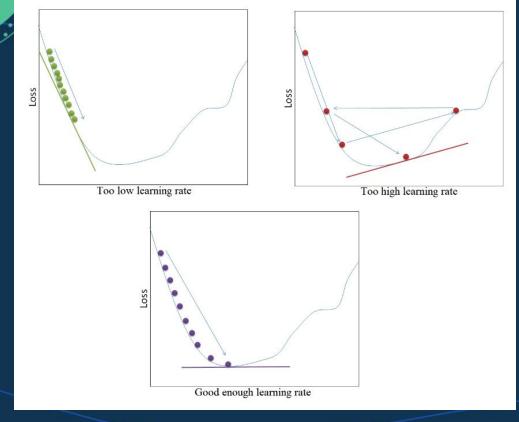
- Parameter tuning is the process of finding the optimal values for a model's hyperparameters to improve its performance.
- Hyperparameters are settings that are not learned from the data but are set before training the model.
 - Examples of hyperparameters in linear regression include the learning rate, regularization strength, and the number of iterations.





Source: Analytics Vidhya





Source: Analytics Vidhya



Which of the following best describes the purpose of simple linear regression?

- 1. To classify data points into distinct categories
- 2. To estimate the relationship between two continuous variables
- 3. To cluster data points based on their similarities
- 4. To predict the probability of an event occurring



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What does the coefficient of determination (R-squared) represent in a linear regression model?

- 1. The average squared difference between predicted and actual values
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- 3. The slope of the regression line
- 4. The intercept of the regression line



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Which Python library is commonly used for implementing machine learning algorithms, including linear regression?

- 1. NumPy
- 2. Pandas
- 3. Matplotlib
- 4. Scikit-learn



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What is the primary difference between supervised and unsupervised learning?

- 1. Supervised learning uses labelled data, while unsupervised learning uses unlabelled data
- 2. Supervised learning is used for classification, while unsupervised learning is used for regression
- 3. Supervised learning requires more computational resources than unsupervised learning
- 4. Supervised learning is more accurate than unsupervised learning



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Which of the following is an example of a regression problem?

- 1. Predicting the sentiment of a movie review (positive or negative)
- 2. Identifying the species of a flower based on its features
- 3. Estimating the price of a house based on its size and location
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Questions and Answers





Thank you for attending







