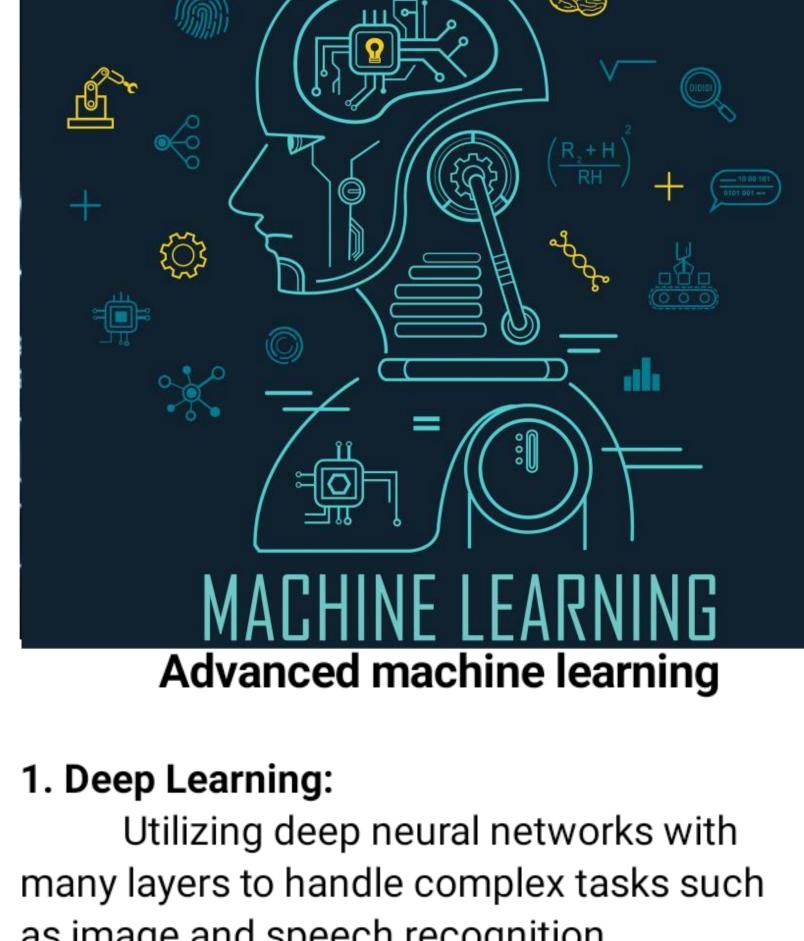
DATA ANALYTICS AND COGNOS



used in autonomous systems like self-driving cars and game playing. 3. Transfer Learning:

Leveraging pre-trained models on

large datasets to improve performance on

a specific task with limited data.

generation and style transfer.

4. Generative Adversarial Networks (GANs): Creating new data that resembles existing data, commonly used in image

5. Natural Language Processing (NLP):

language text, enabling applications like

Analyzing and generating human

Modeling and predicting sequential

sentiment analysis. 6. Time Series Analysis:

data, often used in finance, weather

7. Unsupervised Learning:

forecasting, and demand forecasting.

chatbots, language translation, and

Discovering patterns and structures in data without labeled examples, including techniques like clustering and dimensionality reduction.

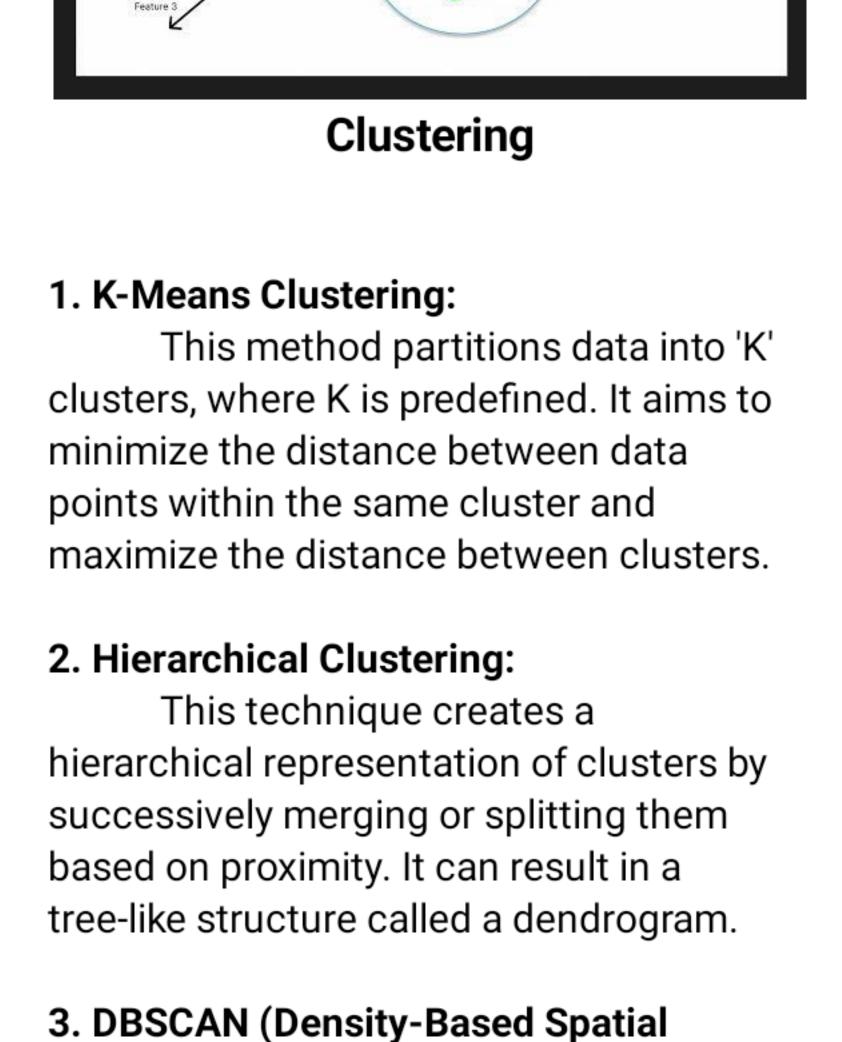
CLUSTERING

Cluster 1

Cluster 2

Cluster 3

Clustering:



Clustering of Applications with Noise):

on the density of data points. It can find

this method starts with individual data

points as clusters and iteratively merges

them based on proximity until a stopping

5. Gaussian Mixture Models (GMM):

clusters of arbitrary shapes and is robust

DBSCAN identifies clusters based

4. Agglomerative Clustering: Similar to hierarchical clustering,

criterion is met.

cluster shapes.

Host factors

Previous infection

Underlying health

Viral variant factors

T

Label

100H

50K

100K

2016

0

Tooltip

vaccine

Antigenic mismatch with

Increased transmissibility

1. Data Collection:

time-stamped information.

conditions

Immune compromise

Genetic polymorphisms

Old age

7. Mean-Shift Clustering:

to noise.

GMM assumes that data points are generated from a mixture of Gaussian distributions. It estimates the parameters of these distributions to identify clusters. 6. Spectral Clustering: Spectral clustering uses the eigenvalues and eigenvectors of a

similarity matrix to perform clustering. It

can be useful for non-convex and complex

Mean-shift is a density-based

clustering technique that seeks modes in

varying shapes and sizes. Vaccine Distribution and

Demographic factors

living together

High levels of circulating virus

High levels of vaccine uptake

High levels of herd immunity

Vaccine effectiveness

Positive effect on VE

Negative effect on VE

Unknown effect on VE

(a) vaccine Distribution

Gather historical data on vaccine

distribution and adverse effects, including

Close proximity of people

January James

Vaccine access factors

Which vaccine used

Timing between doses

Heterologous prime-boost

Cost-benefit decisions by

Limited access to vaccines

national vaccine bodies

Number of doses

Immune factors

(neutralization)

T cells

High antibody titres

Quality of antibody

adverse effects data:

the data density. It can find clusters of

2. Data Preprocessing: Clean and preprocess the data, handling missing values and outliers, and ensure it's in a suitable format for time series analysis. 3. Exploration: Visualize the data to identify any apparent patterns or trends. You can use tools like line plots, scatter plots, or autocorrelation plots. Pages Filters Sheet 1 Segment Corporate Marks

7. Model Evaluation:

networks.

6. Model Training:

 Feed supply Urber ja padatione Population growth Trinickel and Tebago EMI_e DTF3 coverage (%) Death lates 40 tie nographics Leonomics Valeur (downer) Formula amploy want was or ce NAWE EVB Other Scheding VAPIR

G

Analyze the model's forecasts and identify any hidden patterns or insights that emerge from the data.

and potential adverse effects. 9. Interpretation:

Fine-tune your model or explore

alternative methods if the initial results are

components. This can be done using methods like seasonal decomposition of time series (STL). 5. Model Selection: Choose an appropriate forecasting model based on the characteristics of your data. Common models include ARIMA (AutoRegressive Integrated Moving Average), Exponential Smoothing methods, or machine learning algorithms like LSTM (Long Short-Term Memory)

Decompose the time series data into its underlying components, such as trend, seasonality, and residual (error)

2017

4. Time Series Decomposition:

2018

Month of Order Date

(b) Time Series Forecasting

2020

Evaluate the model's performance using appropriate metrics like Mean Absolute Error (MAE), Mean Squared Error (MSE), or Root Mean Squared Error (RMSE) on the test data.

C

1990

2000

Femak education

Water access

Population age structure

2023-10-10

2010

Split your data into training and

testing sets, and train your selected

forecasting model on the training data.

1990 ვები 1980

Afgharistan

100

-Hungary

Malcives:

Mean time in acrool (years).

10. Refinement:

not satisfactory.

100 100 (Socine coverage (S) DTP3 cove 309 (%) 0-64 0.64 1.400 10 化 0.66 1:00 0.78 20 0-63 0.87 0-92 1-00 --- DTP1 ---- MCV FUR 表於形 ĐOG DTP3 1980 D 85 Female education (years) Note: access to enable (%) 60 45 70 201p 1980 1990

Venazuala

(C) Adverse Effects Data 8. Forecasting: Use the trained model to make future forecasts of vaccine distribution



as image and speech recognition. 2. Reinforcement Learning: Teaching machines to make

Project: Covid Vaccines Analysis Phase -2 Advanced machine

