Categorizing documents by topics in Pent analysis => Catagorizing documents by topics in Text analysis typically another using techniques like natural anguage Processing and Machine Learning algorithms. =) These algorithms con analyze the content of documents to determine their main themes or topics based on the words and phrases used. · Common methods Prelude. (1) Polic modellinge It's type of stantistical model for discovering topics that occur to a collection of document of decument of decument of description (LDA) is a popular typic modelling technique that assumes each document is a mixture of topies. i con a presence => The topics produced by topic modeling technique

Categorizing documents by topics Po Pext analysis => Catagorizing documents by topics in Text analysis typically anolves using techniques like natural hanguage Processing and Machine Learning algorithms. =) These algorithms com analyze the content of documents to determine their moin themes or topics based on the words and phrases used. · Common methods Prelude. (1) Topic modelling. It's type of startistical model for discovering topics that occur to a collection of document ) Latent Desichlet allocation (LDA) is a popular topic modelling technique that assumes each document is a moture of topies. conta words o presence =) The topics produced by topic modeling techniques are clusters of Simelar words.

-) It is used for finding semantic stanting of test body 2) Latent Décidlet algorithm étasatively condates 15 à flesible generative probabliste model. 3) It 95 a unsupervised model. DA assumes each document is a set of topics, and that each word for the document is attributable to one of the document's topics." 3) Intelally, the model assigns woods to the documents > LDA Hexates to Propriese the assignment of woods to topics based on peobabilities until it Converges to a stable solution. > During 9 texations, It adjusts the assignment of words to topics based on perobabilities. ) The Output of LDA is hypeally a set of topics, Each represented by a list of words that one most lakely to occur within that topic also for each document, the probability of belonging to Vasiones topiss.

Clustering Algorithms.
These algorithms group documents
90to clusters based on heir Amilosety Po
terms of word usage.
K-means clustering is commonly used for
this propose
Don Lancon America lance
Deep Learning Appearables.
neural networks, can also be used for document
Categorization.
eg: Recussent newal networks (RNNs)
Convolutional newal networks (CNN)
COLOR DE LA COLOR

EDGE

Determining sentiments in Tool analysis It is the process of computationally Palentifying and categoxizing opinions from piece of text, and determine whether the coster's attitude towards a particular topic or the product Es positive, negative, neutral. For this we's foll we are using following approaches:-· Levicon - Based approaches These methods use sentement lexicons, which are dictionaries containing words labeled with their sentement (positive, negative or neutral). he sentement of a document les détermencel by aggregating the Sentements of Its constituent words. · Machine Leaving-based approaches Machine loaning techniques, lette classification algorithms (eg: Support Vector machines, Naive Bayes, News)

can be used to classify tent into sentiments. ⇒ By applying these techniques, Sentiment analysis can provide Valuable Posights to public opinion customer feedback, Docal media trends and more. It helps to organizations to make decessions and condenstand the sentement of their Gaining Insights in Text Analysis 3 Gaining Insights from tent analysis involves extracting meaningful enformation, patterns, and knowledge from textual data to make decisions and to understand underlying trends Note: You can write the Steps to left cycle also

Hadan Emeritaria
Hadoop Ecosystem
) Apache Hadoop is an opensource frame
work Potended to make Poteration with big data
Casier
Started Vascue delegation of Statestand and
=> Hadoop ecosystem 93 a platform ox suit
aled be store 1600 as a confess to sole the tradat
which provides vacaus services to solve the bydak
peroblems.
→ It Portudes Various Commercial tools and
Solutions.
Following one the components that collectively from a
11 18 0 15
Hadep System
* HDFS (Hadeop * HBASE * Zookaspee
Distabuted Bysters) (No Sal DB) (Coluster Marrye)
* YARN-Yet omther * Squap * Ooxie Resource Negotiator) (Data technyc) (Job shedily)
* MAPREDUCE - * Flume
( log control delais)
(Query based processy (Machine Learning Liberries)
data Services)

HDFS Name Hode master EDGS Map reduce HDFS & the premary or major Component of Hedoop ecosystem and 9t's Desponsible for Strong large datasets of strentwed and constructured data across Various nodes and thousy maintaining the metadata is the form Name hode that quides the datanade as well as stores netde Data node act as slave nodes are marrly used to store the data. Map Reduce · MapReduce 83 a processing technique and a program model for distributed computing based on java. 9 It can process Vast amount of data % parellel across distributed computing envisonments

The map reduce algorithm contains 2 Emportant 1. Map 2. Reque. Map stage. The map/mapper's job is to process the Popul data. Generally, % p data is & the form of file or directory and is stored to the Hedoop file System (HDFS). The 9/p file is passed to the mapper function line by line The mapper processes the data and creates Several Small chunks of data. Reduce Stage This stage is the combination of shuffle The reduces's job 15 to pricess the data.

that comes from mapper After processing et produces a new sot of output, which will be is the HOFS Map reduce phases. (i) Map phase. The first phase es map reduce es known as map, draing which the dataset file is divided Poto multiple splits. each split is passed Poto its constituent Records as Key-value paix. The key is usually the oxdenal position of Record, and the value is actual record. The mapper processes each key-value paiss as per user logic and further generates a Key-Value pair as of.

(2) Combine phase Generally, the ofp of the map function is handled directly by the Reduce function. How ever map tasks and reduce tasks mostly sun ove different nodes. But, there is a combine function that Summarize a mapper's ofp before it gets by the Reduces. Combin N N Note # # # is govolved, a partition divides op from mapper esto partitions b/w sauce

Shuffle and Sort Duxing first stage of reduce task, of from all positioners is copied across the n/w to nodes sunning the reduce task. This is known as Shuffling. Then It automatically groups and sorts according to the Keys. So that 0/p contains a sorted list of all ep keys and their Values with Some Keys appearing together Reduce is the final stage of the Reduce fast. The reducer cor! Summarize 9t's Poput and will emet the of without making any changes.

EDGA Example Rednie. Mapper. Pactition Popul files This-1 This is -1 This is an apple anam-1 apple apple-1 red-1 Yed-1 Apple 15 Apple-1 Apple is Red 9n Glor > Color Ded-1 Color 1 10-1 color-1 color

Apache Hadop Youn, A 95 called yet another resource Negotrator. It is on apprade to Map reduce present in Hadoop It is efficient resource manager that helps Support applications Such as HBase, Spark and the Yarn am work pasellely with HDFS HBASE 98 a top-level aparhe peroject water on jave which fulfills the needs to read and waste data is Realtine

It provides a Bimple Poterface to the destributed data. It can be accessed by Spacke thre, Aprile. + House architecture has 3 mais components Hmaster, Region Server, Zookeeper Zookeeper. cleent Hmastes Ragion Serva Region Sever Region Server (Region) (Region) Region (Region) Region (Region.) HDFS

The Poplementation of Master Server Po HBase 93 Homaster. It is a process to which regions are assigned to region Server as well as- DDL Coreate, delete table operations It manitor all xegron Server Region Server HBase Tables are divided hoxizontally by how kere Range Poto Regions. Regions are the base buildings elements of HBase cluster that consists of the distribution of tables and one comprised of coloumn families. Regeon Server rums on HOFS Data Node which is present in Hadoop Cluster.

Pig was basically developed by Jahos which works on a pig later language. Which is query based language. Similar to Sal. It is a platform for structuring the data flow, processing and analyzing large data sets. Prog does the work of executing commands no the background, all achivities of map reduce. one take one of After processing prog stores the Result Po HDFS. ) With the help of Sac methodology and Potesface thre performs Reading and withing 2) It's quary language is called Hal Hive query. => It is hearly scalable as It alous real-time processing and Batch processing. Also all the Sal datatypes are supported by Hive, thus making the query processing lasies. =) Firster to quay processing frame works, HIVE too comes with two components: JDBC Drivers and HME Command line. = JDBC, along with the ODBC drivers work on establishing the data storage. Deemissions me Connection cohereas HIVE Command line helps on processing of queries. Mahout peropodes an environment for Gesting machine learning applications which ose scalable. Machine learning algorithms allows us to build self-loaning machine

that evalve by objectly without being explicitly programmed. Based on user behaviour data porteens and past experiences of makes Proportant future decisions le com call et a descendant of AI. Mahout does? ) collaborative littering. Mahart mênes user behaviour, their patients and their characteristics based on that et paedicts and make recommendations to the users The typecal use case is e-commone 1) Clustering It againges a similar group of data tagether like asticles can contain blogs, news

Deseach papers etc. ii) Classification It means classifiling and Categorizing data esto various sub departments like axticles com categorized Eto blogs, news, Casay, Research papers etc. IV) Frequest Pleas set missing Hose Mahaut cheeks, which objects are likely to be appearing together and make Suggestions, of they are missing lor eg. cell phone and cover are bought together an general so of you Search for a cell phone, of all recommen you cover and cases.

## **Analytics- Technology and Tools: In-Database Analytics**

### What is In-Database Analytics

- In-database analytics is a technology that allows data processing to be conducted within
  the database by building analytic logic into the database itself. Doing so eliminates the
  time and effort required to transform data and move it back and forth between a
  database and a separate analytics application.
- An in-database analytics system consists of an enterprise data warehouse (EDW) built
  on an analytic database platform. Such platforms provide parallel
  processing, partitioning, scalability and optimization features geared toward analytic
  functionality.
- In-database analytics allows analytical data marts to be consolidated in the enterprise data warehouse.
- Data retrieval and analysis are much faster and corporate information is more secure.
- in-database analytics streamlines the analytics process, enhances performance, reduces complexity, and enables real-time or near-real-time insights generation.
- Companies use in-database analytics for applications requiring intensive processing for example, fraud detection, credit scoring, risk management

### Common examples of in-database analytics solutions include:

- SQL-Based Analytics: Utilizing SQL queries with advanced analytical functions directly within the database system.
- 2. Database-specific libraries: Some databases offer libraries or extensions for machine learning, statistical analysis, and predictive modeling.
- 3. Integrated Analytics Platforms: Specialized analytical platforms or appliances that tightly integrate analytics and database capabilities for high-performance analytics.

# SQL Essentials

#### 1 Joins

- Joins in SQL are powerful operations used to combine rows from two or more tables based on related columns between them.
- They enable the retrieval of data from multiple tables simultaneously by establishing relationships between these tables.

There are different types of joins in SQL:

#### 1.Inner Join:

- Returns rows when there is a match in both tables based on the join condition.
- SELECT \* FROM table1 INNER JOIN table2 ON table1.column = table2.column;

#### 2. Outer Join:

- Returns all rows when there is a match in either the left or right table. If there is no match, NULL values are included for columns from the opposite table.
- SELECT \* FROM table1 FULL JOIN table2 ON table1.column = table2.column;

#### 3. Left Join:

- Returns all rows from the left table and matching rows from the right table. If there is no match, NULL values are included for columns from the right table.
- SELECT \* FROM table1 LEFT JOIN table2 ON table1.column = table2.column;

### 4. Right (Outer) Join:

- Returns all rows from the right table and matching rows from the left table. If there is no match, NULL values are included for columns from the left table.
- SELECT \* FROM table1 RIGHT JOIN table2 ON table1.column = table2.column;

#### 5.Self Join:

- When a table is joined with itself, typically used when the table contains hierarchical data or references to itself.
- SELECT e1.name, e2.name FROM employees e1 INNER JOIN employees e2 ON e1.manager\_id = e2.employee id;

### 2.Set Operations

- Set operations in databases are used to perform operations like union, intersection, and difference on the result sets of SQL queries.
- These operations allow data professionals to combine and manipulate data in various ways.

These set operations are handy for various scenarios in data analytics:

- 1.Data Integration: When combining data from multiple sources, UNION and UNION ALL help merge datasets with or without duplicates.
- 2.Data Validation: INTERSECT can be used to check for overlapping records between different datasets, ensuring data consistency.
- **3.Data Cleansing:** EXCEPT or MINUS can identify data discrepancies or missing records between two datasets.
- **4.Data Manipulation:** Set operations enable data professionals to filter and manipulate datasets in complex ways based on set theory principles.

### Primary set operations:

### 1.UNION:

Combines the result sets of two or more SELECT statements into a single result set. It removes duplicates by default.

SELECT column1 FROM table1

UNION

SELECT column1 FROM table2;



### 2.INTERSECT:

Returns rows that appear in both result sets of two SELECT statements.

SELECT column1 FROM table1

**INTERSECT** 

SELECT column1 FROM table2;

## 3.EXCEPT or MINUS: 146 2921102 Statistim mont stab grandmoo redW moltstgath statist

Returns distinct rows from the first SELECT statement that are not present in the second SELECT statement.

SELECT column1 FROM table1

3. Data Cleansing: EXCEPT of MINUS can identify data discrepancies or missing T930X3

SELECT column1 FROM table2;

# In-Database Text Analysis

- In-database text analysis refers to performing text processing, search, and analysis directly within a database system.
- It involves using the database's capabilities to handle and analyze textual data, enabling various text-related operations without needing to extract data to external tools or platforms.
- This approach is particularly beneficial for managing and analyzing large volumes of textual data efficiently.

### Here are key components and techniques involved in in-database text analysis:

- Full-Text Search: Many database systems offer built-in full-text search capabilities.
   These functionalities allow users to perform keyword-based searches, find specific phrases or words within text fields
- Text Indexing: Databases can create indexes specifically optimized for textual data, enabling faster search and retrieval operations on large volumes of text.
- Text Processing Functions: Database systems may provide functions or extensions for text processing tasks, such as tokenization (splitting text into tokens/words), normalization (converting text to a standard form),
- 4. Text Mining and Analytics: In-database text analysis can include mining insights from text data, such as identifying trends, patterns, or associations within textual information.

dr.pokoninica gina.

# Data Privacy and Ethics

 Data privacy and ethics are fundamental aspects of handling, managing, and utilizing data responsibly.

### **Privacy Landscape**

# 1. Protection of Personal Information:

Data privacy refers to the protection of sensitive and personally identifiable information (PII) of individuals. This includes names, addresses, social security numbers, health records, financial information, etc.

### 2. Legal Compliance:

Adherence to data privacy laws and regulations, such as the GDPR (General Data Protection Regulation) in the European Union or CCPA (California Consumer Privacy Act) in California, which outline rules for collecting, storing, processing, and sharing personal data.

## 3. Consent and Transparency:

Ensuring individuals are informed about how their data is collected, used, and shared. Obtaining explicit consent before collecting and processing their data.

### 4. Data Security Measures:

Implementing robust security measures like encryption, access controls, data anonymization, and regular security audits to protect against unauthorized access, breaches, or data leaks.

# **Rights and Responsibilities**

Rights and responsibilities in data privacy and ethics are crucial aspects that both individuals and organizations need to understand and uphold.

### Rights:

- 1. **Right to Privacy:** Individuals have the right to control their personal data, including how it's collected, used, stored, and shared.
- Right to Access: Individuals have the right to access their own data that's held by organizations and understand how it's being used.
- Right to Correction: Individuals can request corrections or updates to inaccurate or outdated personal data.
- Right to Erasure (Right to be Forgotten): Individuals can request the deletion or removal of their personal data under certain circumstances, especially if it's no longer necessary or if consent is withdrawn.
- Right to Data Portability: Individuals have the right to obtain and reuse their personal data for their purposes across different services.
- Right to Consent: Individuals have the right to give informed consent for the collection and processing of their data. Organizations must obtain clear and explicit consent for data usage.

### Responsibilities:

- Data Protection and Security: Organizations have a responsibility to implement robust data protection measures, ensuring the confidentiality, integrity, and security of individuals' data against unauthorized access, breaches, or misuse.
- Compliance with Regulations: Organizations must comply with data protection laws and regulations applicable to their operations, including GDPR, CCPA, and other regional or industry-specific regulations.

- Transparency and Accountability: Organizations should be transparent about their data practices, informing individuals about how their data is used and handled. They must also be accountable for their data handling practices.
- 4. Data Minimization: Collect and retain only necessary and relevant data. Avoid excessive collection or storage of personal information that isn't essential for business purposes.
- Ethical Use of Data: Use data ethically and responsibly, avoiding discriminatory practices, biases, or unethical exploitation of personal information.
- Respecting Individuals' Rights: Respect individuals' rights regarding their data, including providing access, facilitating corrections, honoring deletion requests, and ensuring data portability.

## **Emerging Technologies in Data Privacy**

- Homomorphic Encryption: Allowing computations to be performed on encrypted data without decrypting it, preserving data privacy during computations.
- Zero Trust Architecture: Operating on the principle of "never trust, always verify,"
  where access controls are continuously evaluated based on various factors like device
  health, user behavior, etc.
- Privacy-Preserving Technologies: Differential privacy, federated learning, and secure multi-party computation, enabling data analysis while preserving individual privacy.