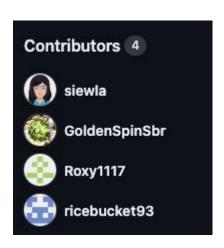
## **Assignment 2**

# Article Classification using Supervised and Unsupervised Learning Approaches

Chi Yuan, Kai Siang, Roshini, Siew La

### **Basics**

- **Github Repo:** <a href="https://github.com/siewla/MCSD2123-assignment-2">https://github.com/siewla/MCSD2123-assignment-2</a>



### **Task Distribution**

#### **Data Preparation:**

Data-Mining and Dataset Cleaning to build Document Class Dataset - Chi Yuan

Perform Exploratory Data Analysis - Chi Yuan

#### Part A:

K-Nearest Neighbours (train model and perform algorithm evaluation) - Siew La

Decision Trees (train model and perform algorithm evaluation) - Siew La

Random Forest (train model and perform algorithm evaluation) - Kai Siang

Support Vector Machine (train model and perform algorithm evaluation) - Kai Siang

Neural Networks (train model and perform algorithm evaluation) - Kai Siang

Thorough Comparison and Evaluation - Siew La

Conclude the Part A findings - Siew La

#### Part B:

K-Means Clustering (perform algorithm evaluation) - Chi Yuan

Hierarchical Agglomerative Clustering (perform algorithm evaluation) - Chi Yuan

DBSCAN (perform algorithm evaluation) - Roshini

Mean-Shift Clustering (perform algorithm evaluation) - Roshini

Thorough Comparison and Evaluation - Roshini, Chi Yuan

Conclude the Part B findings - Roshini, Chi Yuan

#### **Summarization:**

Real Case Implementation - Siew La, Chi Yuan

Summary - Kai Siang

Video Presentation - Kai Siang

## **Data Preparation (Data Mining)**

## 1. Data Extraction

Decide to classify articles into 4 categories. (i.e. 0: Healthcare, 1: AI, 2: IoT, 3: Blockchain). Search the articles and "manually mine" the data to avoid complications due to permission issues in different websites.

### 4.BERT Context Embedding

## 3.Data Pre-processing

The summarized texts are then pre-processed (i.e. text standardization, punctuation removal, lemmatizer etc.).

The pre-processed texts are then fed into BERT to be transformed into context-rich embedding to be saved into our custom datasets.

#### 2.Data Summarization

Data summarization is required due to lengthy article, the BERT embedding size is just 768 tokens. Next, the long article are broken down into multiple groups where text ranging between 250-500. GPT is used to enhance and summarize the contexts after the text trimming.

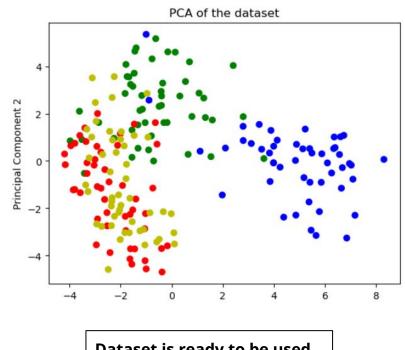
5.Self-Built Dataset is ready

## **Data Preparation (Exploratory Data Analysis)**



768 data points are considered quite small. However, it is still difficult to visualize the embeddings. Thus, PCA is used to reduce the data points to 2.

> **Principal Component Analysis**

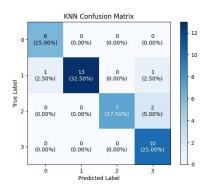


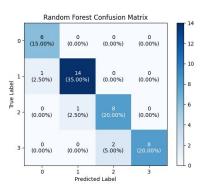
Dataset is ready to be used.

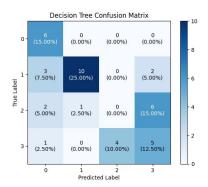
## **Supervised Learning (Models)**

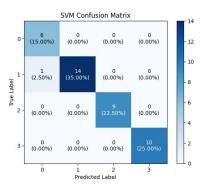
| KNN           | Decision Tree |
|---------------|---------------|
| Random Forest | SVM           |

## **Supervised Learning (Confusion Matrix)**

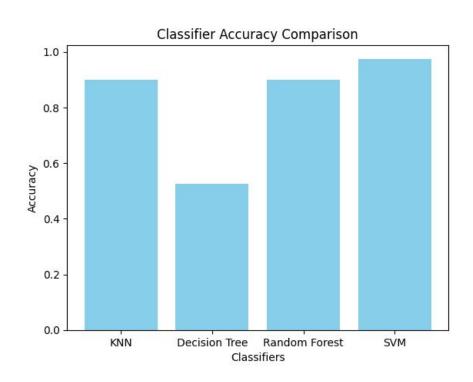








## **Supervised Learning (Accuracy)**



## **Supervised Learning (Summary - KNN, Decision Tree)**

#### **KNN Classifier:**

- **Healthcare (Category 0):** Precision 0.86, Recall 1.00, F1-score 0.92.
- Al (Category 1): Precision 1.00, Recall 0.87, F1-score 0.93.
- IoT (Category 2): Precision 1.00, Recall 0.78, F1-score 0.88.
- **Blockchain (Category 3):** Precision 0.77, Recall 1.00, F1-score 0.87.
- Overall Performance: Generally outperforms the Decision Tree Classifier, especially in predicting IoT instances.

#### **Decision Tree Classifier:**

- **Healthcare (Category 0):** Precision 0.62, Recall 0.83, F1-score 0.71.
- Al (Category 1): Precision 0.92, Recall 0.80, F1-score 0.86.
- **IoT (Category 2):** Precision 0.20, Recall 0.11, F1-score 0.14.
- Blockchain (Category 3): Precision 0.50, Recall 0.70, F1-score 0.58.
- Overall Performance: Performs poorly in predicting IoT instances compared to other categories.

## **Supervised Learning (Summary - Random Forest, SVM)**

#### **Random Forest Classifier:**

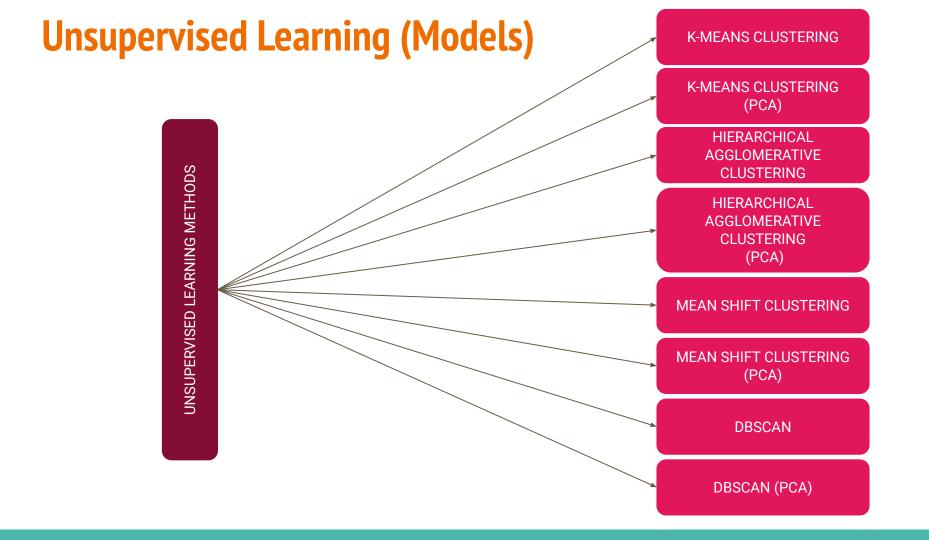
- **Healthcare (Category 0):** Precision 0.92, Recall 1.00, F1-score 0.96.
- Al (Category 1): Precision 0.85, Recall 0.85, F1-score 0.85.
- **IoT (Category 2):** Precision 0.88, Recall 0.93, F1-score 0.90.
- Blockchain (Category 3): Precision 1.00, Recall 0.86, F1-score 0.92.
- Overall Performance: Shows strong and balanced performance across all categories.

#### **Support Vector Machine (SVM) Classifier:**

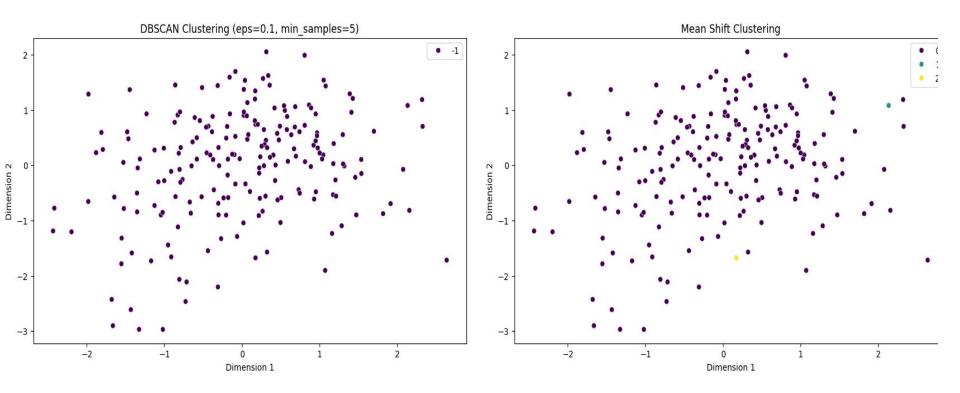
- **Healthcare (Category 0):** Precision 0.92, Recall 1.00, F1-score 0.96.
- Al (Category 1): Precision 1.00, Recall 0.95, F1-score 0.97.
- **IoT (Category 2):** Precision 1.00, Recall 1.00, F1-score 1.00.
- Blockchain (Category 3): Precision 1.00, Recall 1.00, F1-score 1.00.
- **Overall Performance:** Achieves perfect or near-perfect scores across all categories, indicating excellent performance.

## **Supervised Learning (Conclusion)**

- **Best Performer:** The SVM Classifier stands out with consistently high precision, recall, and F1-scores across all categories, indicating robust performance.
- **Challenges:** The Decision Tree Classifier struggles notably in predicting IoT instances, with low precision, recall, and F1-score in this category.
- **Considerations:** Depending on the specific application and priority of metrics (precision, recall, or balanced F1-score), the choice of classifier may vary. SVM is recommended for overall balanced performance, while KNN is noted for its strength in certain categories like IoT prediction. Random Forest also shows strong performance across categories but slightly less so than SVM in precision-recall balance.



## **Unsupervised Learning (Mean Shift Clustering and DBSCAN)**

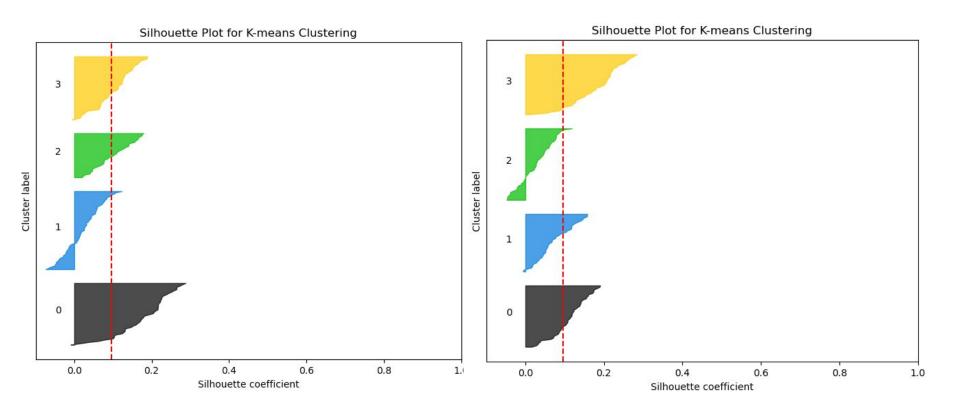


Both algorithms are not suitable for classification of this tasks. PCA could not help the situation.

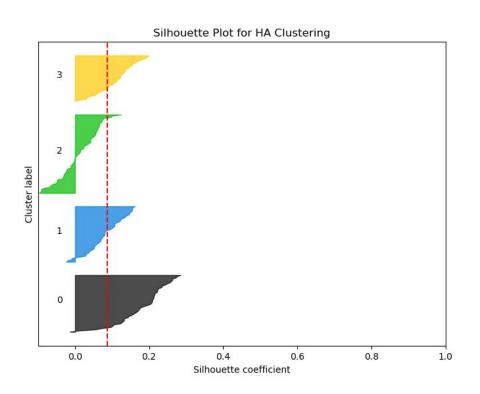
## **Unsupervised Learning (Performance Metrics)**

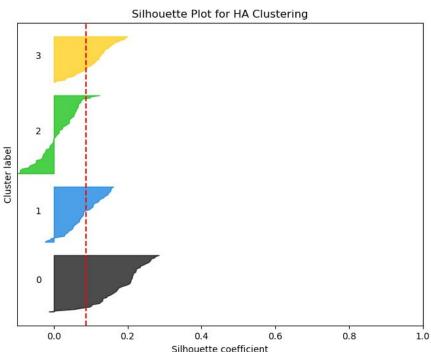
| SCORE                         | K-MEANS<br>CLUSTERING | K-MEANS CLUSTERING with PCA | HIERARCHICAL<br>AGGLOMERATIVE<br>CLUSTERING | HIERARCHICAL<br>AGGLOMERATIVE<br>CLUSTERING<br>With PCA |
|-------------------------------|-----------------------|-----------------------------|---|---|
| Accuracy                      | 0.7000                | 0.7300                      | 0.7250                                      | 0.7250  |
| Silhouette                    | 0.2900                | 0.3278                      | 0.3122                                      | 0.3234  |
| Davies Bouldin<br>Score       | 2.6129                | 2.7318                      | 2.617                                       | 2.8067  |
| Adjusted Rand Index           | 0.4854                | 0.5069                      | 0.49946                                     | 0.4974  |
| Normalized Mutual Information | 0.5264                | 0.5482                      | 0.5479                                      | 0.5423  |
| Cluster Purity                | 0.7100                | 0.7300                      | 0.7200                                      | 0.7200  |

## **Unsupervised Learning (K-Means Clustering)**



# Unsupervised Learning (Hierarchical Agglomerative Clustering)





## **Unsupervised Learning (Comparison)**

#### **Analysis of Each Clustering Algorithm:**

#### 1. K-Means Clustering:

- Strengths: High silhouette score (0.3277), indicating well-defined clusters.
- Weaknesses: High Davies Bouldin index (2.6129), suggesting less distinct clusters.

#### 2. K-Means Clustering with PCA:

- Strengths: Highest silhouette score (0.3277) and good Adjusted Rand index (0.499). High cluster purity (0.72) and accuracy (0.73).
- Weaknesses: Moderate Davies Bouldin index (2.617).

#### 3. Hierarchical Agglomerative Clustering:

- Strengths: High accuracy (0.725) and cluster purity (0.72).
- Weaknesses: Lower silhouette score (0.3122) and moderate Davies Bouldin index (2.617).

#### 4. Hierarchical Agglomerative Clustering with PCA:

- Strengths: High accuracy (0.725), good silhouette score (0.3233), and cluster purity (0.72).
- Weaknesses: High Davies Bouldin index (2.8066).

## **Unsupervised Learning (Conclusion)**

Evaluating the clustering algorithms based on the provided metrics, **K-Means Clustering with PCA** emerges as the best among the existing models. It provides a good balance across the key metrics:

- Highest silhouette score: 0.3277
- Moderate Davies Bouldin index: 2.617
- Highest Adjusted Rand index: 0.499
- **High accuracy**: 0.73
- High cluster purity: 0.72

Nevertheless, it indicates that Unsupervised Learning has poorer performance as compared to Supervised Learning in classifying articles with defined classes.

# Real Case Classification using SVM (Best Classification Model for this study)



insulin production in som



#### What is SDN?

Software-defined networking (SDN) is a softwarecontrolled approach to networking architecture driven by application programming interfaces (APIs). SDN leverages a centralized platform to communicate with IT infrastructure and direct network traffic.

SDN creates and operates a series of virtual overlay networks that work in conjunction with a physical underlay network through the use of software. SDNs offer the potential to deliver application environments as code and minimize the hands-on time needed for managing the network.

Companies today are looking to SDN to bring the benefits of the cloud to network management and deployment. With network virtualization, organizations can achieve greater efficiency through new tools and technology, such as software as a service (SaaS), infrastructure as a service (BasS) and other cloud computing services, as well as integrate via APS with their software-defined network.

SDN also increases flexibility and visibility into network behavior. In a traditional environment, a router or switch—whether in the cloud or physically in the data center—is only aware of the status of network devices next to it. SDN centralizes this information so that organizations can view and control the entire network and devices.

#### What Is a Blockchain?

A blockchain is a distributed database or ledger shared among a computer network's nodes. They are best known for their crucial role in cryptocurrency systems for maintaining a secure and decentralized record of transactions, but they are not limited to cryptocurrency uses. Blockchains can be used to make data in any industry immutable—the term used to describe the inability to be altered.

Because there is no way to change a block, the only trust needed is at the point where a user or program enters data. This aspect reduces the need for trusted third parties, which are usually auditors or other humans that add costs and make mistakes.

Since Bitcoin's introduction in 2009, blockchain uses have exploded via the creation of various cryptocurrencies, decentralized finance (DeFi) applications, non-fungible tokens (NFTs), and smart contracts.

SVM is able to predict class of the articles (e.g. Healthcare, AI, IoT, Blockchain) accurately. However, there is still chance that the SVM wrongly classify between AI and IoT. Simple reasons to it due to closely-related domains. Besides, data loss happened during text trimming and article text grouping into smaller parts due to embedding constraints.

An approach to mitigate the error is to get the average class prediction. For example, an Al article text is broken down in 10 groups. 10 groups could have different predictions, but most of the prediction still biased towards "Al". By averaging the scores, actual predicted class can be attained.

## **Assignment 2 Conclusion**

## Supervised learning Vs Unsupervised Learning in Article Classification Tasks

Supervised learning uses labeled data to train models that can be categorized articles into predefined classes.

Unsupervised learning clusters articles into groups with similar context without labels.

Thus, Supervised Learning achieved better results in this assignment. However, Unsupervised Learning is still useful in clustering larger amount of unlabelled data into groups for further post-processing processes.

## THANK YOU