

Optimising Student Internet Navigation: A Comparative Analysis of Machine Learning Algorithms for Action Prediction

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Table of Contents

Introduction

Problem Definition

Existing Technologies

Proposed Solution

Research Methodology

Data Collection Framework

Determine Best Algorithm

Implementation

Evaluation

Conclusion



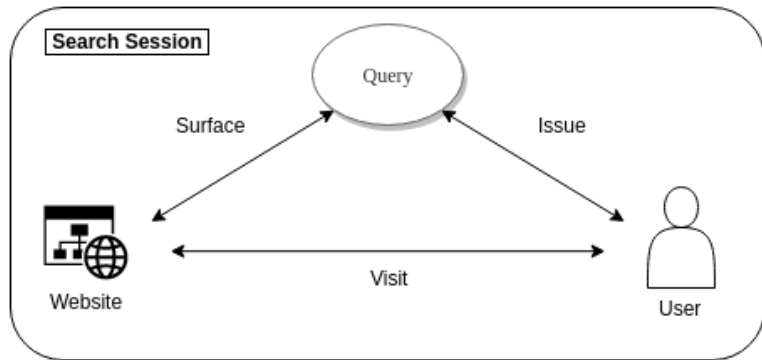


Figure: A web search session as explained in Kim et al. 2012



Problem Definition

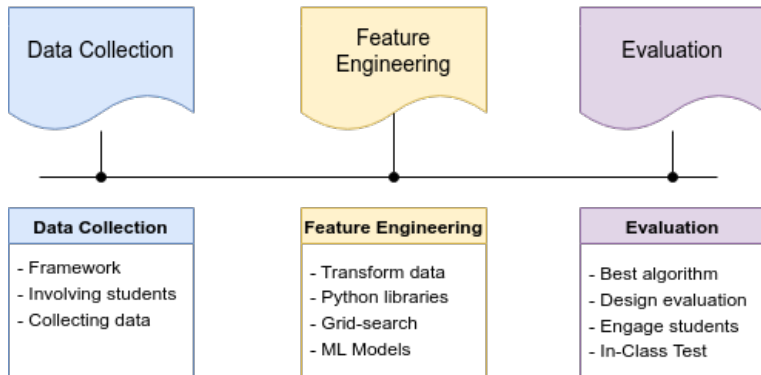
- Students have access to a wealth of online information and educational resources.
- Students require such information to complete assignments and prepare for exams (Tsai 2009)
- Finding useful information online requires a good searching strategy and prior domain knowledge.
- Novice students who lack domain knowledge will struggle (Debowski 2001).
- Keyphrases searched by students in queries across cohorts are not retained and are forgotten over time.



Next Word Prediction

- An application of Natural Language Processing (NLP).
- Predicting the next most suitable word in a sentence based on preceding words (Rathee and Yede 2023).
- More words more accurate the predictions will be (Rathee and Yede 2023).
- Commonly used in mobile devices
- Improves typing experience (Lehmann et al. 2023)
- **Time-Series** Prediction.
- Forecasting future values and trends (Shi et al. 2023).





Data Collection Framework

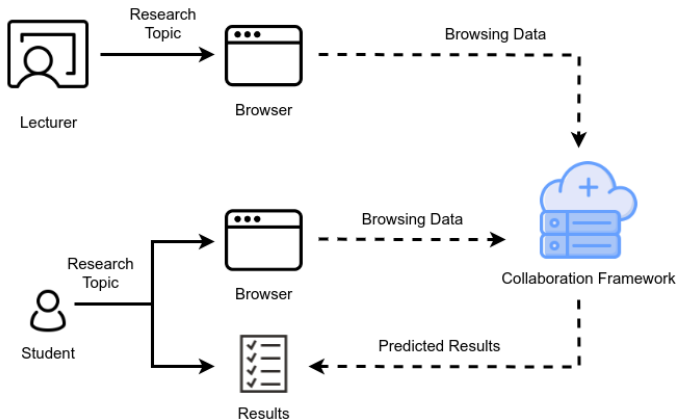


Figure: Collaborative Framework Overview





<https://www.google.com/search?q=machine+learning>



<https://stackoverflow.com/search?q=machine+learning>



https://en.wikipedia.org/wiki/Machine_learning

Table: Extracting Keyphrases

sequence_key	action_data
22fc4e4e_2023_8_24_18	machine learning
22fc4e4e_2023_8_24_18	artificial intelligence
22fc4e4e_2023_8_24_18	data science
22fc4e4e_2023_8_24_18	supervised learning
22fc4e4e_2023_8_24_18	www.techtarget.com
22fc4e4e_2023_8_24_18	unsupervised learning
22fc4e4e_2023_8_24_18	data science



Constructing Sequences (1 Word)

Table: Sequences (1 Word)

sequence_key	action_data
22fc4e4e_2023_8_24_18	machine learning
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Constructing Sequences (3 Word)

Table: Sequences (3 Word)

sequence_key	action_data
22fc4e4e_2023_8_24_18	machine learning
22fc4e4e_2023_8_24_18	artificial intelligence
22fc4e4e_2023_8_24_18	data science
22fc4e4e_2023_8_24_18	supervised learning
22fc4e4e_2023_8_24_18	www.techtarget.com
22fc4e4e_2023_8_24_18	unsupervised learning
22fc4e4e_2023_8_24_18	data science



Metrics Used for Evaluation

Precision:

Precision measures the accuracy of positive predictions.

$$\text{Precision} = \frac{TP}{TP + FP}$$

Recall (Sensitivity):

Recall measures the ability of the model to capture all positive instances.

$$\text{Recall} = \frac{TP}{TP + FN}$$

F1 Score:

The F1 score is the harmonic mean of precision and recall.

$$F1 = 2 \times \frac{\text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$$



Results

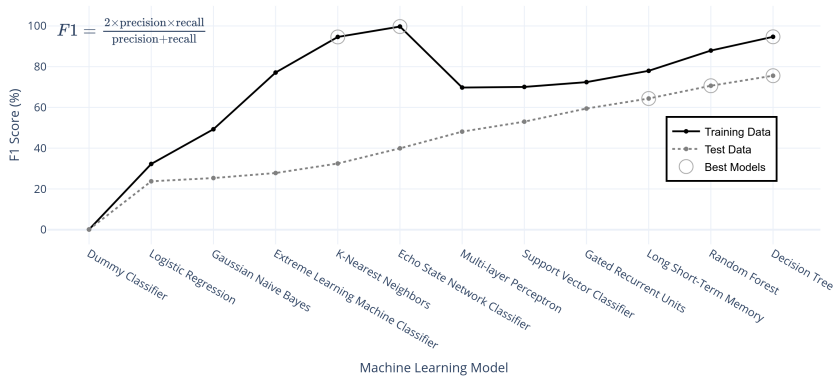
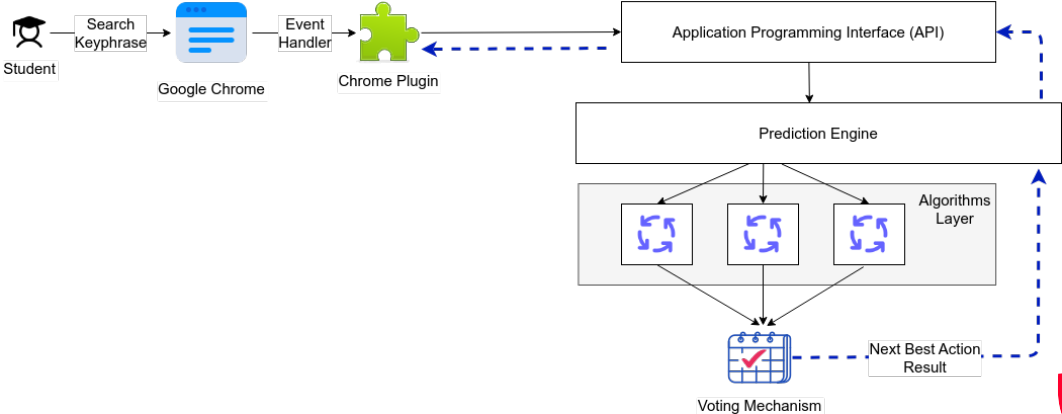


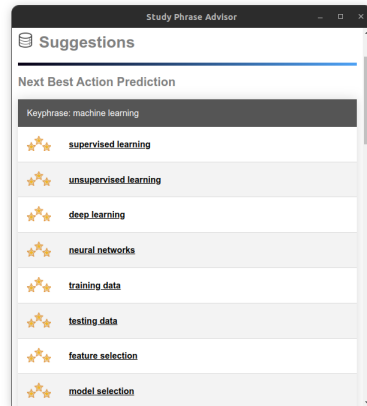
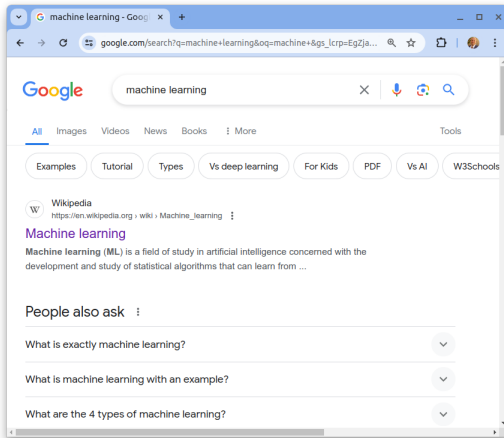
Figure: F1 scores obtained after the grid search.



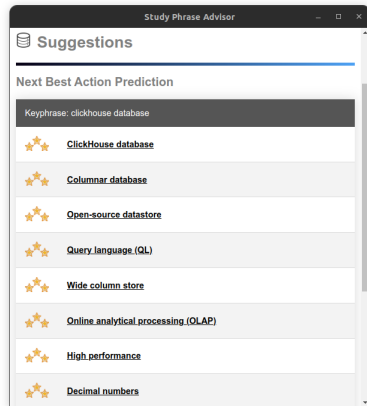
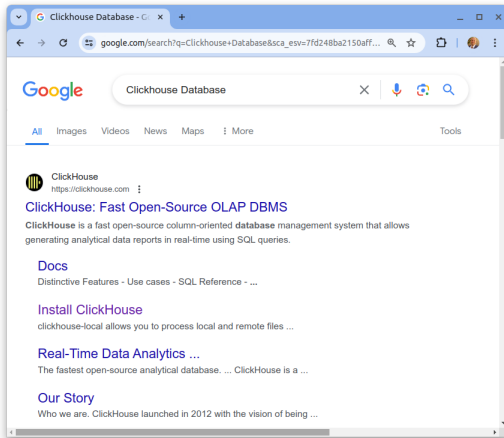
Prediction Engine



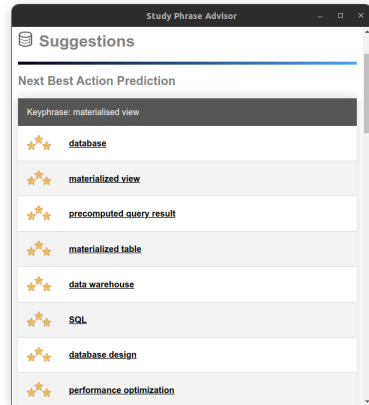
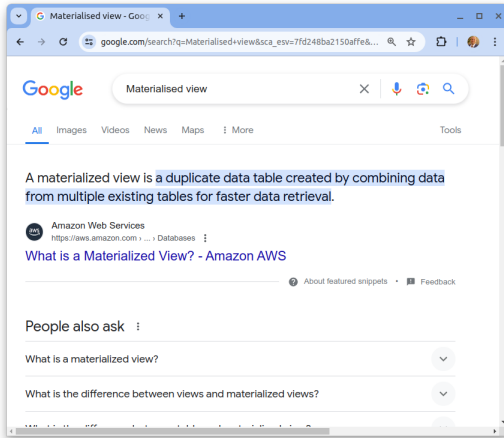
Example 1: Training During a Machine Learning Class



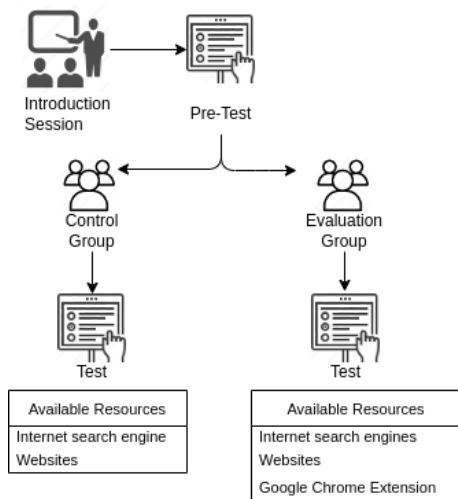
Example 2: Training While Preparing For a Lecture



Example 3: Lacking When Searching Some Keyphrases



In-Class Test design



- Brainstorming session with the subject matter lecturer to design tests.
- Focused on topics not yet discussed during the lectures.
- Closed-ended multiple choice was used to avoid subjective answers.
- Question type included: Acronymns, Problem Solving and Theoretical.
- 6 students acted as the control group while 5 as the evaluation group.



Aggregated Results

Table: Evaluation Results

	Evaluation	Control
Total Students	5	6
Pre-Test Average Score (\bar{P})	2.00%	6.67%
Test Average Score (\bar{T})	84.00%	58.33%
Difference ($\bar{P} - \bar{T}$)	82.00%	51.66%
Std Dev (σ)	19.24	19.41



Pros and Cons

- ✓ Knowledge gain - The evaluation group performed better.
- ✓ Motivation - Students showed a high level of enthusiasm while using the application.
- ✓ Novice Students - Such students will gain advantages from key phrases extracted from modules of previous years.
- ✗ Privacy - Students expressed concerns regarding the data collection system during browsing.
- ✗ Unseen Data - When the system lacks training on a particular topic, the suggestions provided may be ambiguous.



- Data collection, model selection and evaluation of various machine learning algorithms.
- Results showed that the evaluation group performed better.
- The approach provided a pedagogical benefit.
- Future research opportunities include evaluation across diverse educational domains and student populations.



Thank You

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