



Chapter 3: Methodology

BERT-BASED SEMANTIC SIMILARITY OF MALAYSIAN
LEGAL PRECEDENTS

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Chapter 3: Research Methodology

This chapter outlines the methodology used to develop the **BERT-based semantic similarity model** for Malaysian legal precedents. The focus is on the systematic approach taken to enhance legal research efficiency and accuracy.



Research Objectives and Research Questions

Research Objective (RO)

RO1: To investigate existing semantic similarity approaches in legal NLP tasks and identify gaps in the context of Malaysian legal texts.

RO2: To develop, fine-tune, and compute the semantic similarity of Malaysian legal precedents using a BERT-based model.

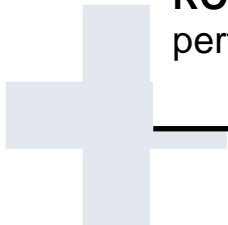
RO3: To evaluate and visualize the model's performance using key evaluation metrics.

Research Question (RQ)

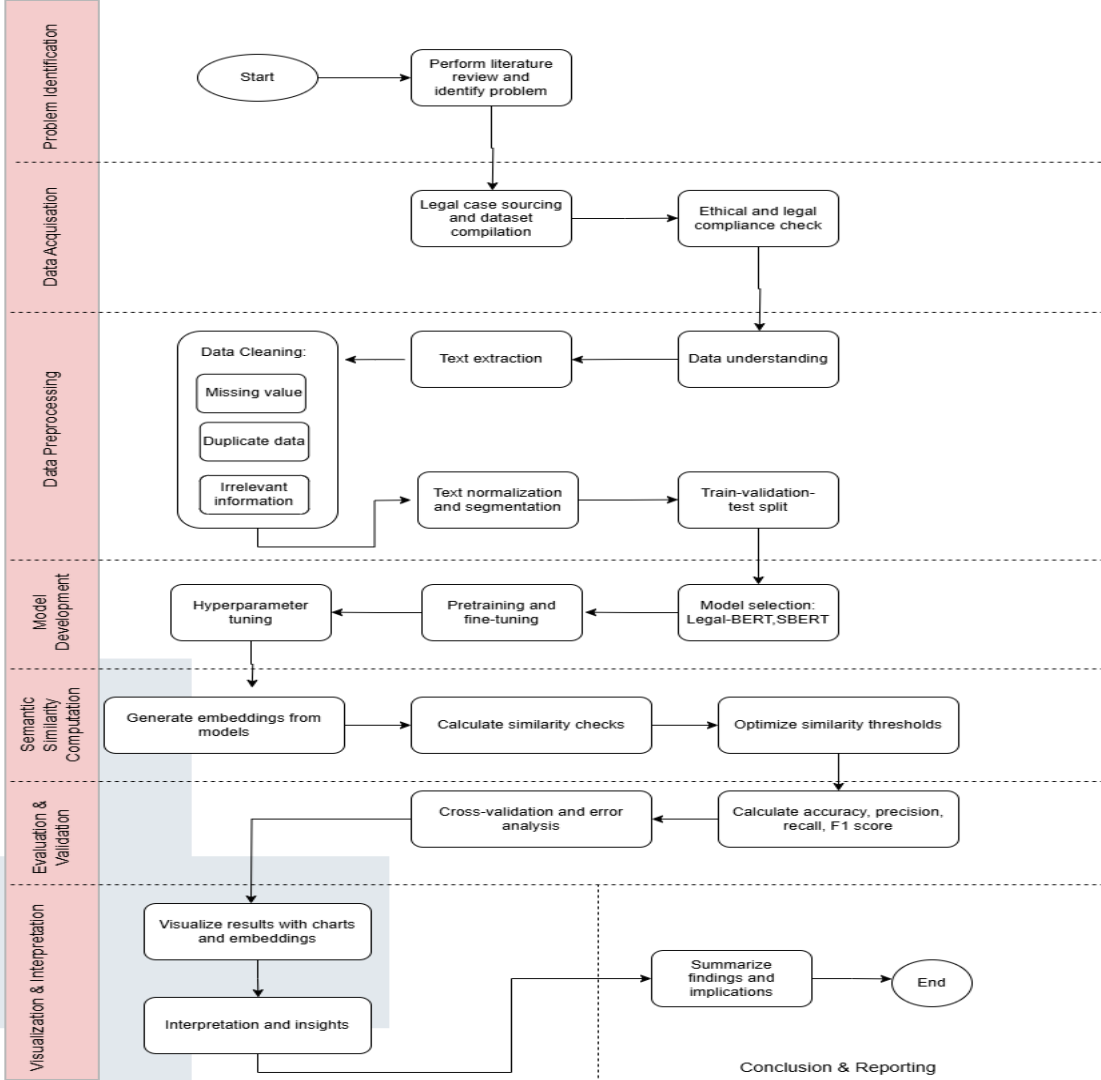
RQ1: What are the challenges and limitations in applying semantic similarity models to Malaysian legal texts?

RQ2: How can a BERT-based model be fine-tuned and applied to compute the semantic similarity of Malaysian legal precedents?

RQ3: How can the model's performance metrics and similarity results be visualized to enhance understanding and interpretation by legal professionals?



Research Framework



PHASE 1

PROBLEM IDENTIFICATION

- Define research problem: Challenges in semantic similarity for legal NLP tasks.
- Conduct literature review: Identify gaps and existing solutions in legal NLP.
- Formulate research questions: Focus on improving semantic similarity measurement in Malaysian legal texts.

PHASE 2

DATA COLLECTION AND UNDERSTANDING

- Legal case sourcing: Download legal documents (court judgments, statutes).
- Dataset compilation: Collect data from LexisNexis, Malaysian court databases.
- Ethics check: Ensure compliance with data usage rights.
- Feature analysis: Analyze the legal features affecting semantic similarity (case type, legal arguments, etc.).

PHASE 3

Data Preprocessing

1. Pre-processing stage:

Text Extraction (Extract text from PDF or other document formats)
Data Cleaning (Remove irrelevant metadata, Eliminate noise)
Normalization (Lowercasing, Punctual Removal, Whitespace Management)

2. Processing stage:

- Sentence Segmentation (Use NLP tools (e.g., SpaCy) for accurate segmentation)
- Feature Extraction (Identify key features within the text (e.g., legal terms, case citations, rulings))
- Handling Imbalance

3. Post-Processing stage (Deep Transfer Learning):

- Train-Test Split (Divide the dataset into training, validation, and test sets)
- Final Clean-Up (final cleaning to ensure the processed data is ready for model training (e.g., removing remaining stopwords, lemmatizing)).

PHASE 4

Model Development

- Model selection
- Pretraining & fine-tuning: Pretrain on a general corpus, fine-tune on Malaysian legal dataset.
- Hyperparameter tuning: Optimize hyperparameters (learning rate, batch size, etc.).

PHASE 5

Semantic Similarity Computation

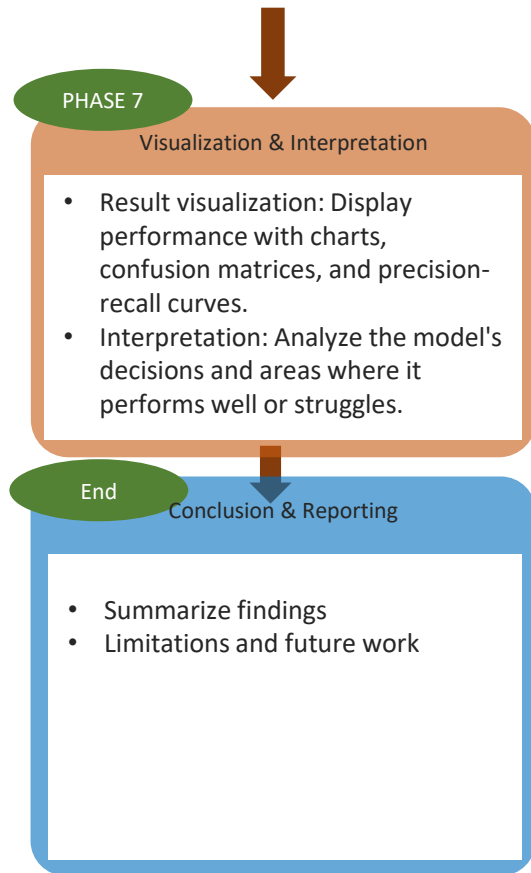
- Embedding generation: Generate embeddings from the fine-tuned model.
- Similarity measurement: Use cosine similarity or other metrics to calculate document similarity.
- Threshold optimization: Set a similarity threshold to classify cases as similar or not.

Phase 6

Model Evaluation

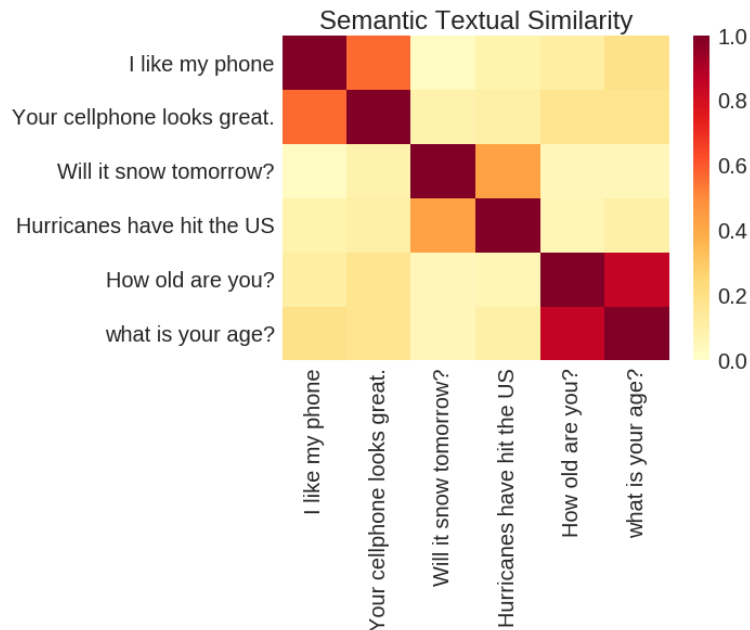
Performance metrics: Evaluate accuracy, precision, recall, F1 score, etc.
Cross-validation: Validate the model on different data splits to check robustness.
Error analysis: Identify misclassifications and potential areas for improvement.

Cont.



Problem Identification

The challenge of **semantic similarity** in Malaysian legal precedents is significant. The motivation behind this research is to enhance legal research efficiency and accuracy, allowing legal professionals to find relevant precedents more effectively.



Dataset	Features	Cases/Instances	Legal Domain	Up-to-date Relevance	Legal Text Representation	Distinctive Characteristics
Malaysian Legal Dataset	60+ (e.g., case type, court, judgment)	5,000+ (selected for training)	Malaysian Law	Focuses on recent cases in Malaysian law	Full case texts, judgments, rulings	Includes citations, legal references, and metadata
LexisNexis Malaysian Case Law	50+ (e.g., citation, case type, year)	10,000+ (subset of 100,000+ cases)	Malaysian law	Covers current legal trends in Malaysia	Full judgments and summaries (PDF and other formats)	High-quality , authoritative content used in legal research
Court Case Corpus (Malaysia)	45+ (e.g., court, case facts, issues)	2,000+ (random sample)	Criminal, Civil, Constitutional	Publicly available recent judgments	Text of judgments, rulings	Focus on common legal issues in Malaysia
Public Malaysian Legal Data	40+ (e.g., court level, case year)	1,000+ (public cases)	Civil, Criminal, Administrative	Available through open government channels	Legal documents, summaries, statutes	Emphasizes open access and government transparency

PHASE 3: Data Processing

Stage	Task	Description
Stage 1: Data Preprocessing	Text Extraction	Extract legal case text from downloaded PDF or other formats (e.g., LexisNexis).
	Data Cleaning	Remove irrelevant metadata, OCR artifacts, and noise from the text.
	Normalization	Convert text to lowercase, remove punctuation, extra spaces, etc.
	Tokenization	Split the text into tokens (e.g., words, sentences, or paragraphs).
Stage 2: Feature Engineering and Segmentation	Text Segmentation	Divide the text into meaningful sections (e.g., facts, issues, judgments).
	Feature Extraction	Extract legal features such as case type, legal terms, and citations.
	Handling Class Imbalance	Apply techniques (e.g., SMOTE, random oversampling) to balance dataset.
Stage 3: Post-Processing	Train-Test Split	Split the data into training, validation, and testing datasets.
	Text Vectorization	Convert text to machine-readable embeddings using models like BERT.
	Threshold Optimization	Set similarity thresholds to classify legal case relevance.

Task	Description
Model Selection	BERT-based model (Legal-BERT, SBERT) for semantic similarity tasks.
Pretraining and Fine-Tuning	Pretrain the model on a generic corpus, and then fine-tune it with the Malaysian legal dataset to adapt it to the domain.
Hyperparameter Tuning	Tune hyperparameters such as learning rate , batch size , number of epochs , to improve model performance.

Task	Description
Embedding Generation	Generate semantic embeddings for the legal documents using the fine-tuned model.
Similarity Measurement	Measure semantic similarity using metrics like cosine similarity , Manhattan distance , or other suitable metrics.
Threshold Optimization	Set an appropriate similarity threshold to classify legal precedents as relevant or irrelevant .

Task	Description
Performance Metrics	Evaluate the model's performance using metrics like accuracy , precision , recall , and F1-score .
Cross-Validation	Perform k-fold cross-validation to ensure model robustness and generalization.
Error Analysis	Analyze misclassifications to identify specific errors or weaknesses in the model's predictions.

Task	Description
Result Visualization	Visualize model performance using charts , graphs , and confusion matrices .
Interpretation of Results	Interpret the model's results to understand why certain cases were deemed similar or dissimilar.
Insights Generation	Generate insights based on the results, including the model's strengths and weaknesses in legal case retrieval.

Research Phase	Research Objectives (RO)	Deliverables
Phase 1: Problem Identification and Literature Review	RO1	Chapter 1 & 2
Phase 2: Data Acquisition and Dataset Preparation	RO2	Chapter 4
Phase 3: Data Preprocessing	RO2	Chapter 4
Phase 4: Model Development and Fine-Tuning	RO2	Chapter 4
Phase 5: Semantic Similarity Computation	RO2	Chapter 4
Phase 6: Model Evaluation and Validation	RO3	Chapter 5
Phase 7: Visualization and Interpretation	RO3.	Chapter 6
Phase 8: Conclusion and Reporting		Chapter 6