1. Problem Definition

• Goal: Classify text messages (emails, SMS, etc.) as spam or ham (not spam).

2. Data Collection

- Use datasets like the **SMS Spam Collection** dataset from UCI Machine Learning Repository or any relevant labeled dataset.
- Format: Typically has two columns: label (spam or ham) and message (the actual text).

3. Data Preprocessing

- Convert labels to binary (e.g., spam \rightarrow 1, ham \rightarrow 0).
- Clean the text:
 - Lowercasing
 - Remove punctuation, numbers, and special characters
 - Remove stopwords (like "the", "is", "in")
 - Tokenization (splitting text into words)
 - o (Optional) Lemmatization or stemming
 - Explanation

1. Tokenization

Meaning: Breaking a sentence or paragraph into smaller parts like words or punctuation.

Example:

```
Text: "I love playing football."

Tokens: ['I', 'love', 'playing', 'football', '.']
```

This helps machines understand and process each word separately.

2. Stemming

Meaning: Cutting the word to its root form by removing prefixes or suffixes, even if the result is not a valid word.

Example:

- "playing", "played", "plays" \rightarrow "play"
- "running" → "runn" (not a real word, but stemmed)

It's fast but sometimes gives rough or inaccurate root words.

3. Lemmatization

Meaning: Converting a word to its proper dictionary root form, considering grammar and context.

Example:

- "playing" → "play"
- "better" → "good"
- "studies" → "study"

More accurate than stemming, but a bit slower.

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4. Feature Extraction (Text Vectorization)

- Convert text into numerical format using:
 - Bag of Words (CountVectorizer)

0	TF-IDF (TfidfVectorizer)
0	(Optional) Word embeddings like Word2Vec, BERT for advanced models
5 Split Your Data	
Divide your dataset into training and testing sets:	
6 Model Selection	
• Choo	se a classification model:

o Naive Bayes (MultinomialNB) — best for text classification

Support Vector Machine (SVM)

o Logistic Regression

- o Random Forest / XGBoost
- o (Advanced) Deep Learning: LSTM, BERT

7. Model Evaluation

- Use metrics:
 - Accuracy
 - o Precision, Recall, F1-Score (important for imbalanced datasets)

- Confusion Matrix
- Validate with cross-validation if needed

8. Prediction and Testing

- Use the trained model to classify new messages.
- Evaluate its real-world performance.

9. Save the Model

• Use joblib or pickle to save the trained model and vectorizer:

```
python
CopyEdit
import joblib
joblib.dump(model, 'spam_classifier.pkl')
joblib.dump(vectorizer, 'vectorizer.pkl')
```

10. Deployment(local in vs_code)

• Build a web or mobile app using Flask, Django, or Streamlit to input new messages and show classification results.