# RETRIEVAL AUGMENTED CODE GENERATION AND SUMMARIZATION

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### What?

We propose a framework for automatic source code generation and summarization, in which we:

- Constructed a comprehensive legal dataset that consolidates various Vietnamese legal documents, including laws, decrees, and circulars
- Developed REDCODER, a hybrid system that combines Retrieval-Augmented Generation (RAG) with pre-trained models like PLBART and GraphCodeBERT to improve the accuracy and contextual relevance of generated code and summaries

## Why?

 Source code generation and summarization are essential in software development, aiding productivity and code comprehension. While developers frequently seek related examples during coding, most existing models depend solely on pre-trained architectures without leveraging external knowledge, limiting performance on complex tasks. By integrating retrieval with generation, our approach improves contextual understanding and delivers more accurate, developer-like outputs

## **Overview** Generation Retrieval Generate output via Retrieve top-k related **PLBART** using retrieved code/summary with SUMMARY SCODE-R context RETRIEVED CODE/ SUMMARY OUTPU' **Encode code & summary** into vector DB in pair

# **Description**

## 1. Indexing

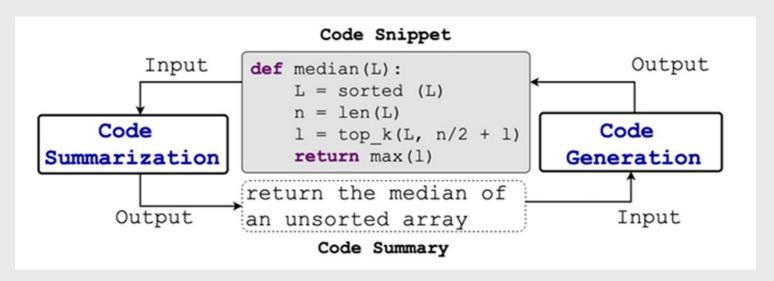
 Code and summary pairs, along with standalone entries, are encoded using pretrained models like CodeBERT and GraphCodeBERT, then indexed in a vector database. This setup enables REDCODER to efficiently retrieve semantically relevant context for both generation and summarization, reflecting real-world developer practices.

### 2. Retrieval

• Given a code snippet or summary as REDCODER input, uses SCODE-R—built on Dense Passage Retriever—to retrieve top-k relevant code or summaries from the vector database. It leverages pretrained encoders like CodeBERT and GraphCodeBERT to measure semantic similarity, enhancing input meaningful context before generation.

# 3. Face-Track Matching

 The retrieved context is combined with the original input and passed to SCODE-G, a generation module based on PLBART. This enhanced input allows the model to generate accurate code or summaries, with support for both single and paired retrieval formats, improving fluency, correctness, and contextual relevance



**Figure 1**. Example input/output for code generation and summarization tasks.

