





# Comparing Natural Language Embeddings for Libc Functions as Rich Labels

Bachelor defense

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# Outline

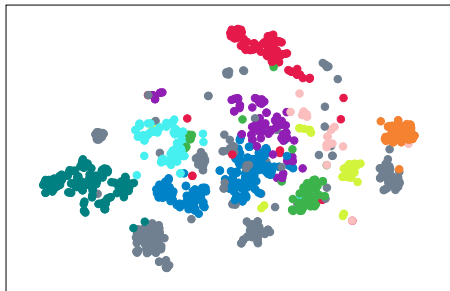
Motivation & Research Objective

Methodology

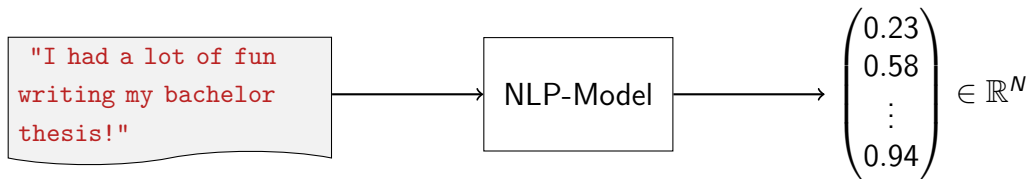
Results

Limitations

Conclusion & Future Work

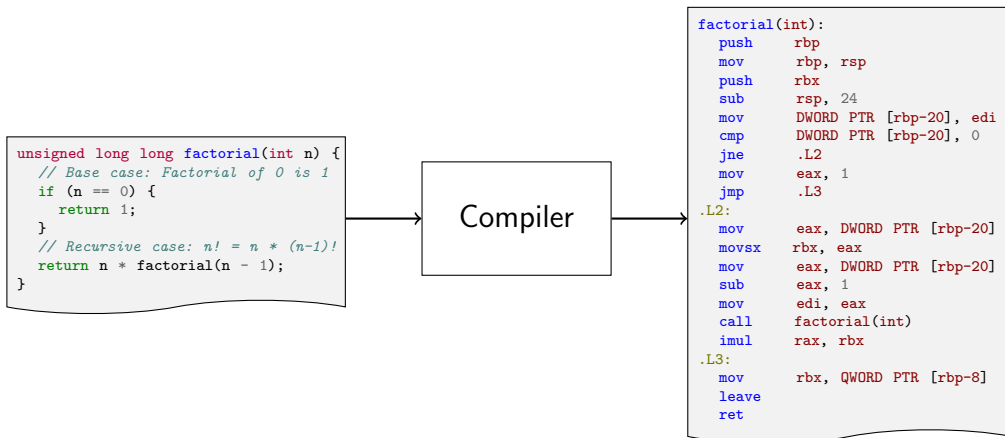


# Motivation



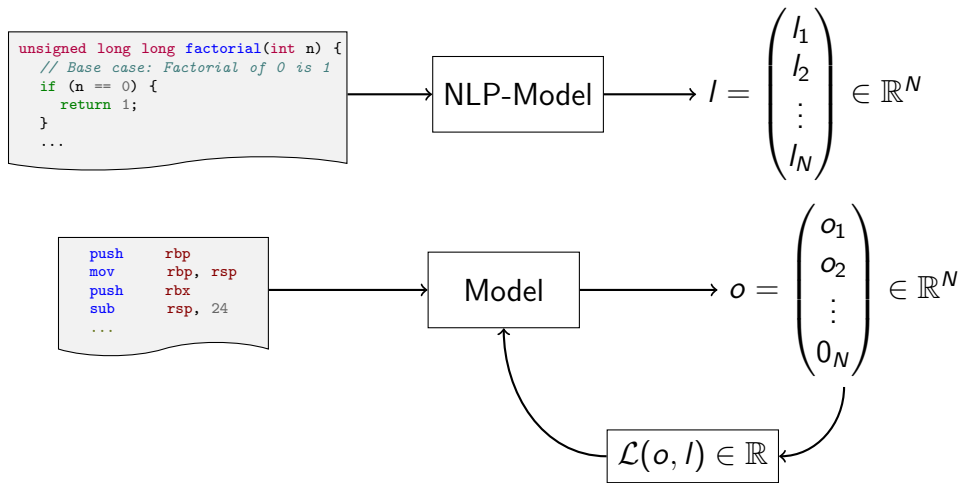
- ~> Encoding natural language was a huge factor in recent nlp advancements
- ~> Information described as a vector can be used in many downstream task
- ~> That motivates encoding binary code and describing them as a vector
- ~> That motivates using NLP tools to encode binary code

# Motivation



⇒ Compiler removes important information in natural language

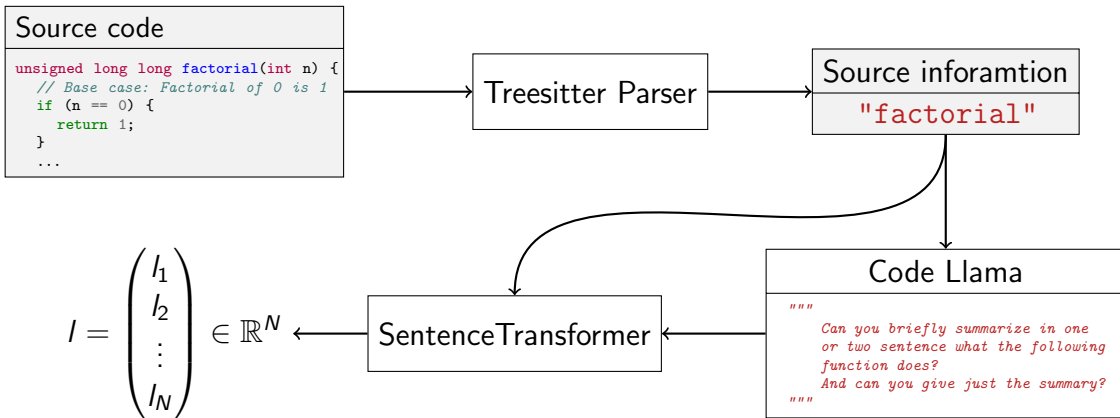
# Motivation



# Research Objectives

- ▶ Compare different approaches generating an Embedding with NLP tools
  1. Embed function names with SentenceTransformer
  2. Embed function comments with SentenceTransformer
  3. Embed Code-Llama code summaries with SentenceTransformer
- ▶ Compare NLP approach to the existing Code2Vec Model
- ▶ Propose a new way comparing embedding spaces

# Architecture





# Expert Survey

"execl"

1. j1l
2. exp10l
3. exp2l
4. expm1l

☐ Yes

☐ No

"fmaximum\_numl"

1. fminimum\_magl
2. fminimuml
3. fminimum\_mag\_numl
4. fminimum\_numl

☐ Yes

☐ No

Figure: Positive exmaple

Figure: Negative exmaple

Ergebnisse der Expertenbefragung				
Strategie	Code-Llama-Erklärungen	Funktionsnamen	Funktionskommentare	Code2Vec
Score	0.596	0.532	0.433	0.321

# Embeddings space comparison

# Embeddings space comparison

# Evaluation with T-SNE

HALLO

HALLO

# Discussion