

NeSy is alive and well: A LLM-driven symbolic approach for better code comment data generation and classification

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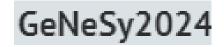


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Task Introduction

- Classification of code comment dataset in C language
- Initially proposed at the Information Retrieval for Software Engineering (IRSE) workshop at the Forum for Information Retrieval Evaluation (FIRE)
- Evaluation of usage of LLMs to augment dataset synthetically
- RQ1: Can symbolic learning help LLMs generate more diverse data samples?
- RQ2: Can symbolic learning overcome the output limitation of LLMs for data generation?
- RQ3: Can symbolic learning make LLM data generation more explainable?





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Dataset - Seed

- 11452 code comment pairs in C language
- Comments labeled Useful or Not Useful
- 7063 Useful samples
- 4389 Not Useful samples
- Dataset curated from real GitHub projects

#	Comment	Code	Label
1	/* uses png_calloc defined	$/*$ uses png_calloc defined in pngriv.h*/	Useful
	in pngriv.h*/	PNG_FUNCTION(png_const_structrp png_ptr)	
		<pre>{ if (png_ptr == NULL info_ptr == NULL)</pre>	
		return;	
		<pre>png_calloc(png_ptr);}</pre>	
2	/* serial bus is locked be-		Not Useful
2	fore use */	<pre>static int bus_reset () /* serial bus</pre>	Not Oseiul
	fore use 7	is locked before use*/	
		{ update_serial_bus_lock (bus * busR); }	
		<pre> { update_serial_bus_lock (bus * busk); ; }</pre>	
3	// integer variable		Not Useful
	,,, 0	int Delete_Vendor; // integer variable	





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Dataset – Augmentation by Examples

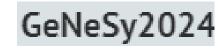
- Generation from examples of synthetic data with ChatGPT 3.5
- 1000 samples generated, 421 unique samples retained
- 411 labeled Useful, 10 labeled Not Useful
- Limit on sample diversity and output token for generation

- Here are the lines with the labels "Useful" or "Not Useful" based on whether the comment \Box \Box \neg matches the code:
- 1. "char letter = getchar();", "Read a character from the standard input." Useful
- 2. "int factorial = calculateFactorial(10);", "Calculate the factorial of a number." Useful
- 3. "printf(""Hello, World!\n"");", "Print the message 'Hello, World!' to the console." Useful
- 4. "Int distance = calculateDistance(x1, y1, x2, y2);", "Calculate the distance between two points in 2D space." - Useful
- 5. "int max = getMaxValue(array, size);", "Find the maximum value in an array." Useful
- 6. "int result = addNumbers(10, 20);", "Add two numbers and store the result in result." Useful
- 7. "Int index = linearSearch(array, size, target);", "Perform a linear search on an array to find a target value." Useful
- 8. "for (int i = 0; i < size; i++) {", "Loop through an array of size elements." Useful
- 9. "if (array[i] == target) {", "Check if the current element is equal to the target value." Useful
- 10. "printf(""Target found at index %d\n"", i);", "Print the index where the target value is found." Useful
- 11. "break;", "Exit the loop early if the target is found." Useful
- 12. "}", "End of the if statement." Not Useful (Comment doesn't match the code)
- 13. "int length = strlen(string);", "Get the length of a string named string." Useful
- $\pm4.\,$ "char ch = 'A';", "Declare a character variable with the value 'A'." Useful
- 15. "int square = num * num;", "Calculate the square of a number." Useful

G Regenerate

16. "if (x == y) {", "Check if the values of x and y are equal." - Useful

Send a message









Semantic Decomposition

- Algorithm that breaks down meanings of concepts into less complex concepts
- Used to generate ruleset of 17 rules from C guidebook (21st Century C: C Tips from the New School)
- Rules are prompted into ChatGPT 3.5 to re-learn coding etiquette and generate better synthetic

data

🌆 You

I will teach you the syntax of a programming language. The syntax will be stated as rules. You will follow these rules strictly to generate correct lines of code. Rule 1: The smallest individual unit of a program is called a token. Rule 2: Tokens are either keywords, identifiers or variables. Rule 3: A keyword must belong to the list: auto, double, int, struct, break, else, long, switch, case, enum, register, typedef, char, extern, return, union, const, float, short, unsigned, continue, for, signed, void, default, goto, sizeof, volatile, do, if, static, while. Rule 4: An identifier can only have alphanumeric characters(a-z, A-Z, 0-9) and underscore(_). Rule 5: The first character of an identifier can only contain alphabet(a-z, A-Z) or underscore (_). Rule 6: Identifiers are case-sensitive in the C language. For example, name and Name will be treated as two different identifiers.

Rule 7: Keywords are not allowed to be used as Identifiers.

Rule 8: No special characters, such as a semicolon, period, whitespaces, slash, or comma are permitted to be used in or as an Identifier. Example of valid identifiers: total, avg1, difference_1. Example of invalid identifiers: \$myvar, x!y.

Rule 9: A variable has a data type (which can be one of the following: char, int, float, double, void), a name and a value.

Rule 10: A variable should be declared and assigned a value. Example: int marks = 10. Rule 11: After creation and assignment, the value of a variable can be changed. Rule 12: A valid line of code is a collection of tokens that adhere to the above rules.

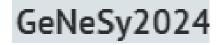
Using the above rules, generate a correct line of code.

🚳 ChatGPT

Certainly! Here's a correct line of code following the provided rules:



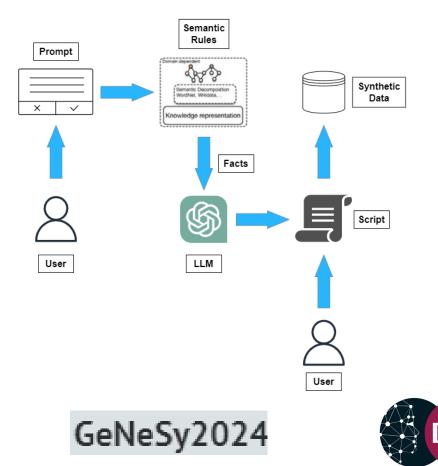
ChatGPT can make mistakes. Consider checking important information.





Proposed NeSy Framework

- Prompting ChatGPT with semantic rules = more controlled data generation
- Tour de force: prompt LLM to output Python script and generate data abiding by ruleset
- Mechanism open to human validation for more control over generated data







Dataset – NeSy Augmentation (1)

- Formal rules allow LLM to generate data in the correct code-comment-label format
- LLM generates more diverse samples
- LLM learn difficult coding principles like correct variable assignment and naming convention
- The power of LLM labeling the data is retained

២ You

Can you generate 100 lines using the same csv format and by applying all the above rules? Please try to vary the examples in the code lines (different keywords, different data types), the comments (level of detail, length) and the labels (Useful or Not Useful)

ShatGPT

Certainly! Here's a snippet of a CSV file with 100 lines, as requested:

Copy code

Line of Code,Comment,Label

float average_score = 75.5; // Declaration of a floating-point variable for average int user age = 25;,// Declaration of an integer variable to store the user's age,Usef double calculateRectangleArea(int length, int width);// Function prototype for calcul struct Node { int data; struct Node* next; };// Structure definition for a basic link const float speed_of_light = 299792458.0;,// Constant representing the speed of light enum TrafficLightColors { Red, Yellow, Green };// Enumeration representing different void swapIntegers(int* a, int* b);// Multi-line comment describing the purpose of the char student_grade = 'B'; // Character variable storing the grade of a student,Useful volatile int hardware_pin_state = 1; // Volatile variable representing the state of #define SQUARE(x) ((x) * (x)) int squared value = SQUARE(5); // Macro for calculating int is_valid = 1; // Single-line comment indicating the initialization of a boolean int count = 0; // Not Useful: int count = 0;,Not Useful float pi_value = 3.14159; // Approximate value of Pi,Useful long total_seconds = 86400; // Total seconds in a day,Useful // Comment explaining the purpose and usage of the printMessage function void printMessage() { printf("Hello, World!"); }// Useful // Multi-line comment detailing the characteristics of a Point structure struct Point { int x; int y; };// Useful

Please note that I've provided a spinnet here due to space constraints. You can continue the pattern for

Message ChatGPT...

ChatGPT can make mistakes. Consider checking important information.







Dataset – NeSy Augmentation (2)

You

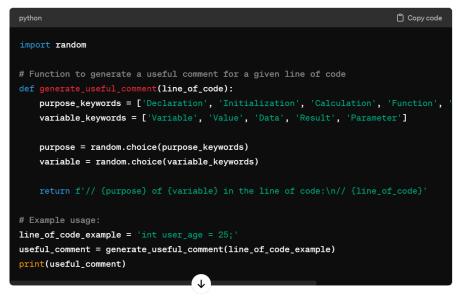
Can you generate the Python code you're using to generate a Useful comment for a line of code?

- Script generation enables human validation on rulebuilding logic and data generation mechanism
- Script is parametrizable and can control number of generated samples and labels
- 5000 new samples generated, evenly split between Useful and Not Useful classes

Mic Innia

ChatGPT

Certainly! Here's a Python function that generates a useful comment for a given line of code:



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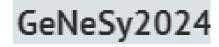


Experimental Setup (1)

- Feature engineering: code + comment in each pair vectorized as vectors of size 768
- Vectorization model based on Sentence Transformers fine-tuned on CodeSearch data
- Label prediction on augmented datasets using 3 classification models
- Phase 1 dataset (baseline): seed + augmented by examples (11873 samples)
 - 7474 Useful, 4399 Not Useful
- Phase 2 dataset: baseline + augmented by NeSy framework (16873 samples)
 - 9974 Useful, 6899 Not Useful









Experimental Setup (2)

- Phase 1 & 2: SMOTE applied to balance class distribution at 50% each
- Phase 1 & 2: K-Fold Cross Validation with 10 splits and 3 repetitions
- Classification models:
 - Classifier A: Random Forest
 - Classifier B: Neural Network (2 hidden layers; 20 neurons followed by 10 neurons)
 - Classifier C: Voting Classifier (majority vote among 3 classifiers)
 - Classifier A
 - Classifier B
 - Linear SVC









Results (1)

- Voting Classifier and Neural Network neck-and-neck in performance
- Addition of synthetic data does not skew model performance
- Increase in scores over all metrics and all models reflects quality of generated synthetic data

	Useful		Not Useful				
Model	Macro-F1	Precision	Recall	Accuracy	Macro-F1	Precision	Recall
RF	88.922	87.186	90.746	88.691	88.448	90.359	86.636
VC	91.468	90.970	91.984	91.418	91.367	91.900	90.853
NN	91.412	92.017	90.829	91.466	91.518	90.954	92.103

 Table 2 Model performance comparison on the augmented data





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Results (2)

- Augmentation by examples (Neural) and augmentation by NeSy (Symbolic) yield positive increase in performance (around 1% each)
- Increase in performance is no indicator alone of quality of data
- Augmentation by NeSy controls diversity, balance and number of samples

Useful			Not Useful					
Model	Seed	Baseline	Augmented	Increase ¹	Seed	Baseline	Augmented	$Increase^1$
RF	84.727	85.587	88.922	1.038	84.168	85.168	88.448	1.038
\mathbf{VC}	88.133	88.539	91.468	1.033	88.111	88.578	91.367	1.031
NN	88.401	88.489	91.412	1.033	88.664	88.856	91.518	1.029

Table 3 Model Macro-F1 performance increase comparison on seed, baseline and augmented data





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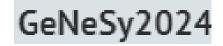


Conclusion

- RQ1: Can symbolic learning help LLMs generate more diverse data samples? Yes
- RQ2: Can symbolic learning overcome the output limitation of LLMs for data generation? **Yes**
- RQ3: Can symbolic learning make LLM data generation more explainable? Yes
- Work consolidates NeSy methodology for synthetic data generation
- Opens the door for solutions to data scarcity problem in different domains









Thank You

Questions







