

Deep Code Comment Generation

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Why Code Summarization?

Code Comprehension

- Comments are often missed, mismatch, outdated, ...

Summarization

 Aims to obtain a reductive transformation from a source text to a summary text through different techniques.

In software development and maintenance, developers spend around **59%** of their time on program comprehension activities*

Xia X, Bao L, Lo D, et al. Measuring program comprehension: A large-scale field study with professionals[J]. IEEE Transactions on Software Engineering, 2017.



Existing Approaches

Information Retrieval Approaches

- Extract natural descriptions from software artifacts, e.g., bug report, Stack Overflow...
- Extract keywords from source code

Limitations:

- ➤ Heavily rely on whether similar code snippets can be retrieved and how similar the snippets are.
- Fail to extract accurate keywords when identifiers and methods are poorly named.



Code Summarization VS NMT

Code Summarization: Bridges gap between Source Code and Natural Language

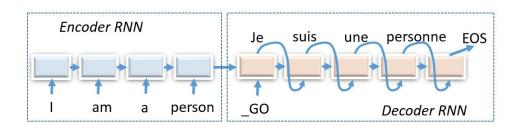
```
public static void sort(Comparable[] a){
  int n=a.length;
  for (int i=1; i < n; i++){
      Comparable v=a[i];
      int lo=0, hi=i;
      while (lo < hi) { ... }
      ...
}
  assert isSorted(a);
}</pre>
Comment:
Rearranges the array in ascending order, using the natural order.

}
```

NMT: aims to translate from one NL language (e.g., English) to another NL language (e.g., Chinese)



NMT Model:





Background: Neural Machine Translation

A deep learning model for the sequence-to-sequence model

Encoder: An RNN that encodes a sequence of words (code)

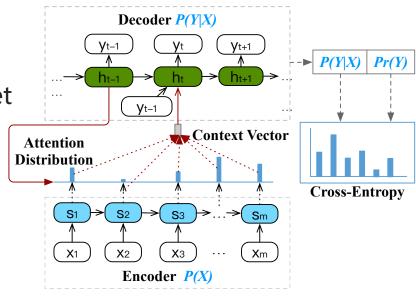
$$s_t = f(x_t, s_{t-1})$$

 Attention: Selects the important parts from the input sequence for each target word.

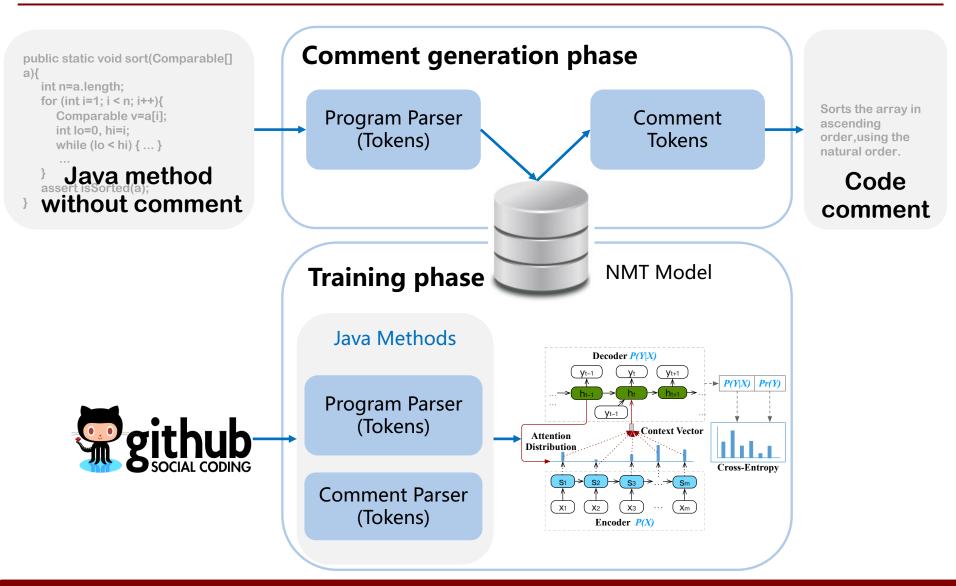
$$c_i = \sum_{j=1}^{m} \alpha_{ij} s_j$$

 Decoder: An RNN that sequentially generates a sequence of words (comments)

$$p(y_i|y_1,...,y_{i-1},X) = g(y_{i-1},h_i,c_i)$$



Intuitive Overview





Program Language (PL)

Natural Language (NL)



Structural

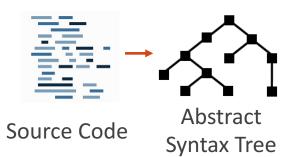
language

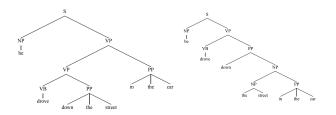
Program Language (PL)

Strong Unambiguous

Natural Language (NL)

Weak
Textual ambiguity





he drove down the street in the car



Program Language (PL)

Structural language

Strong Unambiguous

Vocabulary

Unlimited

Natural	Language
(NL)	

Weak
Textual ambiguity

Limited

- #All Tokens #All # Unique Identifiers Tokens Identifiers 44,378,497 13,779,297 794,711 794,621
- If the vocabulary size is 30,000, 95% identifiers are <UNK>
- If we want the occurrences of (UNK) tokens to be as few as possible, the vocabulary size will increase a lot.



(NL)

Weak

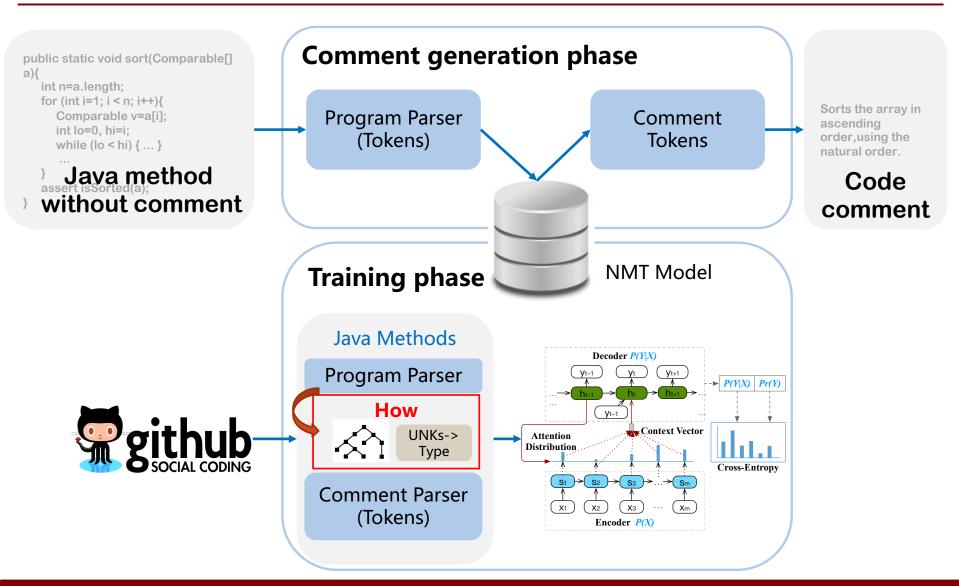
Program Language Natural Language (PL) Strong Structural **Unambiguous Textual ambiguity** language Vocabulary **Unlimited** Limited

Challenges:

- Utilize rich and unambiguous structure information
- Reduce unknown words in source code



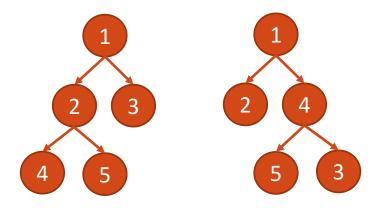
Key Idea





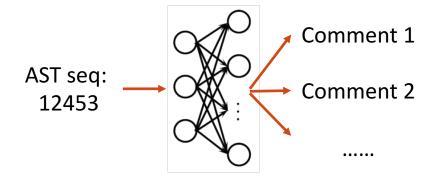
Abstract Syntax Tree with SBT traversal

- Traditional traversal methods, e.g., pre-order
 - Lossy since the original ASTs cannot unambiguously be reconstructed back from them.



Pre-order: 12453

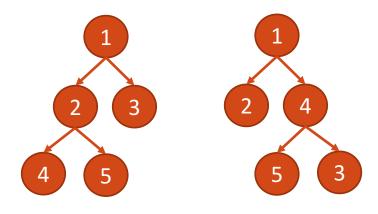
- Different trees (i.e., different Java methods) are mapped into the same AST sequence.
- DNN is confused if there are multiple labels (in our setting, comments) given to a specific input.





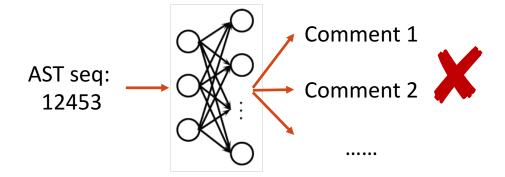
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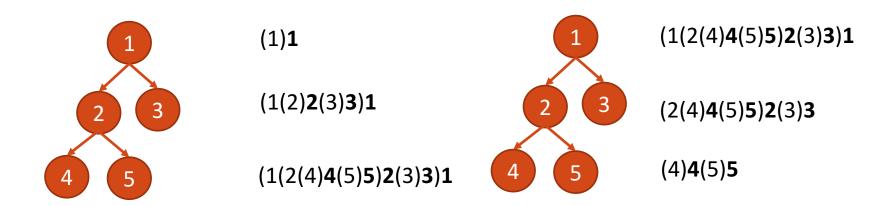




Abstract Syntax Tree with SBT traversal

Structure-based Traversal

Use brackets keeping tree structures





Out-of-Vocabulary tokens

Node representation

- Non-terminal nodes: type
- Terminal nodes: type_value

Vocabulary

- Top 30,000 tokens, including (non-)terminal type, type_value, brackets.
- OOV type_value pairs are replaced by their type.

```
ReturnStatement

MethodInvocation

SimpleName (request)

SimpleName (remove)

SimpleName (id)

SimpleName (extractFor)
```



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SimpleName (request)

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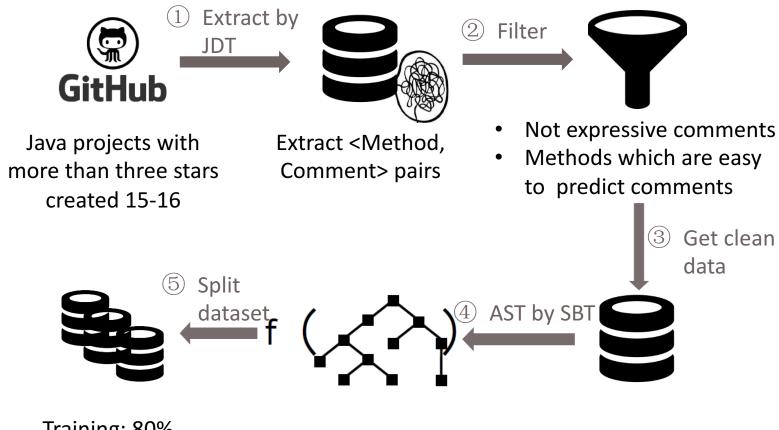
SimpleName (id)
```

```
(ReturnStatement
(MethodInvocation
(SimpleName_request) SimpleName_request
(SimpleName_remove) SimpleName_remove
(SimpleName_id) SimpleName_id
) MethodInvocation
) ReturnStatement
) Block
(SimpleName_extractFor) SimpleName_extractFor

(SimpleName) SimpleName
```



Experiment: Dataset



Training: 80% Validation: 10%

Test: 10%

Generate <AST seq, Comment> pairs

588,108 < Method, Comment> pairs



Results

- RQ1: How effective is DeepCom compared with the state-of-the-art baseline?
- RQ2: How effective is DeepCom to source code and comments of varying lengths?



Results

RQ1: DeepCom VS baselines

Java methods

Baseline: CODE-NN

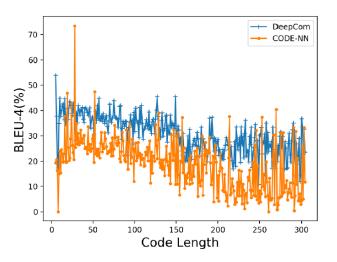
Accuracy Metric: BLEU

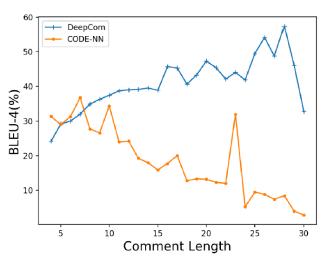
Evaluation results on Java methods		
Approaches	BLEU-4 score (%)	
CODE-NN	25.30	
Seq2Seq	34.87	
Attention-based Seq2Seq	35.50	
DeepCom (Pre-order)	36.01	
DeepCom (SBT)	38.17	



Results

RQ2: Accuracy under different lengths of source code and comments





- (a) BLEU-4 scores for different code lengths
- (b) BLEU-4 scores for different comment lengths
- For most code lengths, the average scores of DeepCom improve those of CODE-NN by about 10%.
- For comments of different lengths, the accuracy of CODE-NN decreases sharply and DeepCom still performs better.



- Exactly correct comments
- Algorithm implementations
- Cases when generated comments are better than humanwritten ones
- API invocations intensive Java methods
- Unknown words in generated comments



Exactly correct comments

- Clear business logic
- Universal code conventions

```
public static byte[] bitmapToByte(Bitmap b){
    ByteArrayOutputStream o = new ByteArrayOutputStream();
    b.compress(Bitmap.CompressFormat.PNG,100,o);
    return o.toByteArray();
}
```

DeepCom: convert Bitmap to byte array
Human-Written: convert Bitmap to byte array

DeepCom: If no profile has been configured, set by default the "dev" profile.

Human-Written: If no profile has been configured, set by default the "dev" profile.



Algorithm implementations

- Java methods usually use similar structures to implement the same algorithm function.
- Capture the correct functionality

```
public static void sort(Comparable[] a){
  int n=a.length;
  for (int i=1; i < n; i++){
      Comparable v=a[i];
      int lo=0, hi=i;
      while (lo < hi) { ... }
      ...
  }
  assert isSorted(a);
}</pre>
```

DeepCom: Sorts the array in ascending order, using the natural order.

Human-written: Rearranges the array in ascending order, using the natural order.



- Cases when generated comments are better than human-written ones
 - Determine something true or not.
 - Developers write interrogative sentences sometimes.

```
public boolean isEmpty(){
    return root == null;
}
```

public boolean contains(int key){
 return rank(key) != -1;
}

DeepCom: Returns true if the symbol is empty. Human-written: Is this symbol table empty?

DeepCom: Checks whether the given object is contained within the given set.

Human-written: Is the key in this set of integers?





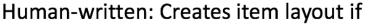
API invocations intensive Java methods

```
public static byte[] bitmapToByte(Bitmap b){
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    return o.toByteArray();
}
```

```
DeepCom: convert Bitmap to byte array
Human-Written: convert Bitmap to byte array
```

```
protected void createItemsLayout(){
   if (mItemsLayout == null){
     mItemsLayout=new LinearLayout(getContext());
     mItemsLayout.setOrientation(LinearLayout.VERTICAL);
   }
}
```

```
DeepCom: Creates item layouts if any parameters
```



necessary

DeepCom: It can be called when the productionly or refresh has ended.

Human-written: Removes a listener that was previously registered with listenFor-Subscriptions.



Unknown words in generated comments

 Not good at learning the method or identifiers names occurred in comments.

```
public FactoryConfigurationError(Exception e){
   super(e.toString());
   this.exception=e;
}
```

DeepCom: Create a new (UNK) with a given

Exception base cause of the error.

Human-written: Create a new

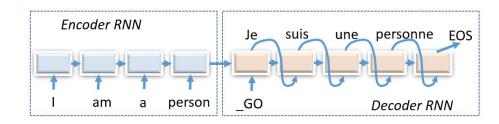
FactoryConfigurationError with a given

Exception base cause of the error.

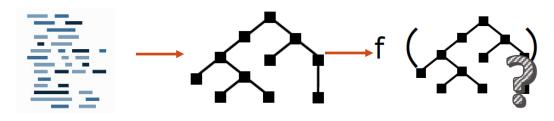


Conclusion

DeepCom: formulate code comments generation task as a machine translation task.



SBT: keep the AST sequences unambiguous



OOV: domain-specific method to deal with OOV

```
private static void UNK(SpringApplication app, UNK source){
    if(!source.UNK("spring.profiles.active")
        &&!System.getenv().containsKey(){
        app.UNK (Constants.UNK);
    }
}
SimpleName
```

Future Work: Domain-specific customizations



Q&A

Thanks



BLEU: the Automatic Metric

- A popular metric for measuring the similarity between two sentences in NMT.
- Range from 0 to 1



DeepCom VS CODE-NN

CODE-NN

 Use an RNN decoder with the attention to generate summaries by integrating the toke embeddings of source code instead of building language models for source code.

DeepCom

Building language model for both source code and comments

