# Summarizing Source Code using a



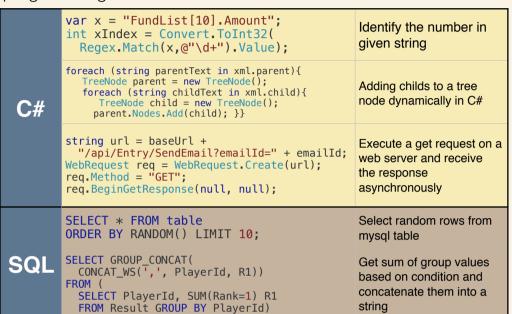
## **Neural Attention Model**



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#### Overview

Our task is to generate high-level summaries of the function of programming source code.



These summaries have many SE applications:







🧑 Code Search 📃 Documentation 📌 Code Navigation

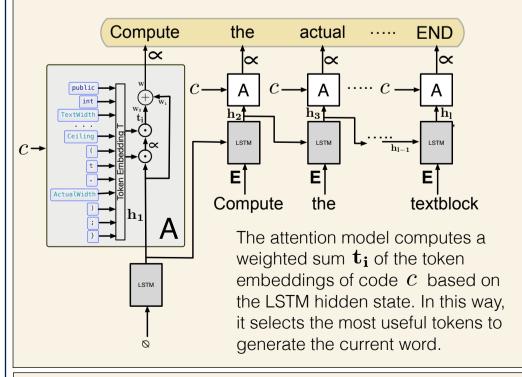
## Neural Attention Model

We use an end-to-end model that jointly performs content selection using an attention mechanism, and surface realization using Long Short Term Memory networks.

We model the conditional next-word probability as:

$$p(n_i|n_1,\ldots,n_{i-1}) \propto \mathbf{W} \tanh(\mathbf{W_1h_i} + \mathbf{W_2t_i})$$

 $\mathbf{h}_{\mathbf{i}}$  is the hidden state of the LSTM cell at the time step i



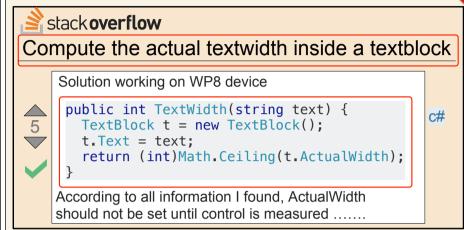
## **Future Work**

- 1. Generation models using tree to sequence methods
- 2. Discovery and explanation of code idioms
- 3. Using language-code models for code synthesis



#### Code Summarization Dataset

We create a new dataset from programming QA websites containing 66K C# and 33K SQL examples.



Code snippets in this dataset are non-trivial:

Loops	> 20%	> 2 Functions	50%	Code	38
Conditionals	> 22%	> 2 Statements	45%	Summary	12

#### Human Annotations

We gather 2 additional references for 200 code snippets for more accurate development and testing.

Data/Code at: <a href="https://github.com/sriniiyer/codenn">https://github.com/sriniiyer/codenn</a>

#### Experiments

Our model beats competitive baselines on summarization metrics and human evaluations.

