

# Information Summarization and Topics generation

DS 595 Natural Language Processing

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## References

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# Introduction

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## Problem Statement

Given the document  $D$ , we want to generate string  $h_D$  such that it summarizes the content in  $D$  with accurate information and concise.



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### Example

$D$ : At the request of the international war crimes court in The Hague, Bosnian Serb police seized the passports of the wife, son, daughter and son-in-law of Radovan Karadzic, the Bosnian Serb leader during the country's war.

$h_D$ : Bosnia: Police Take Passports of Karadzic Family

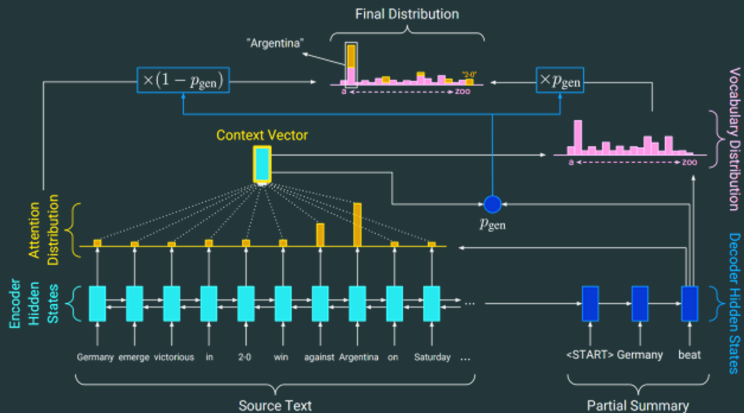
# Methodology

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- As suggested by [Sutskever et al., 2014], RNN model can yield us the promising result but it also yield "artifacts" to the result.

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- Bidirectional RNN [Nallapati et al., 2016] shows the better performance.

# Architecture



Use bidirectional LSTM along with attention to encode and generate context vector [See et al., 2017].

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- What if we combined the method proposed by [See et al., 2017] or [Raffel et al., 2020] with generative model?
- That is, what if we treat [See et al., 2017] or [Raffel et al., 2020] model as the generator and we then create discriminator on top of that?

- NYT news dataset



- NYT news dataset
- >100K entries of title, topic, abstract, keywords
- Problem: Some punctuation might contains semantic value

- Metric of evaluation: BLEU[Papineni et al., 2002] and ROUGE-I[Lin, 2004]