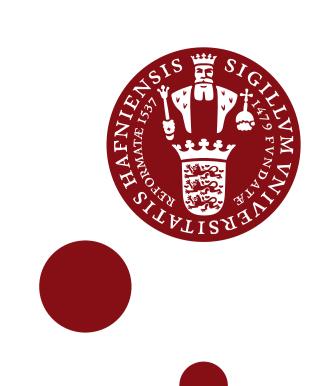


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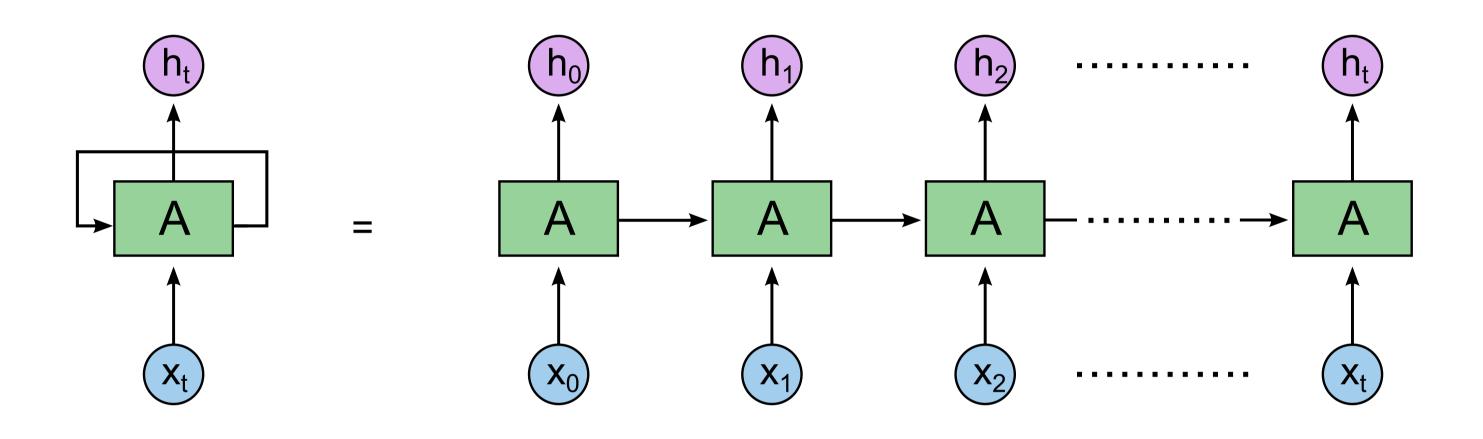


## Deep Learning Relevance: Creating Relevant Information (As Opposed to Retrieving It)

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What if Information Retrieval (IR) systems did not just retrieve relevant information that is stored in their indices, but could also *understand* it and *synthesise* it into a *single* document?

First step in this direction: use deep learning to emulate "learning the semantics" of all indexed relevant information for a query and compose a new document conveying those semantics.



1. Use Long Short Term Memory (LSTM) Recurrent Neural Networks (RNN) to learn relevant information on a character level (3 layers, 512 neurons).

Distributional Semantics Hypothesis: "You shall know a word by the company it keeps"

J. R. Firth

2. Learn only the *distributional semantics* of the indexed relevant information to reduce learning noise.

## **METHODOLOGY**

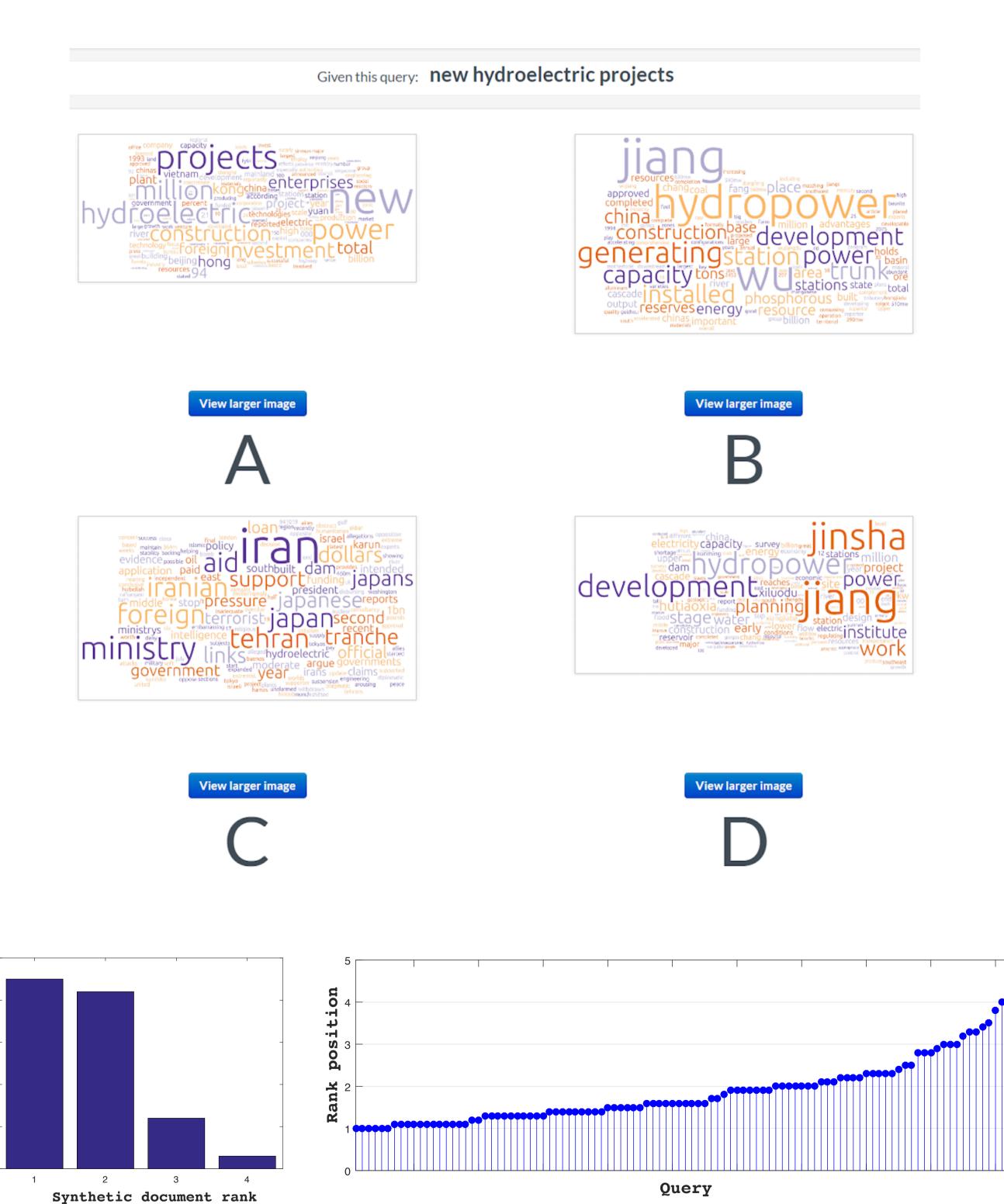
Input: a query and its relevant documents

- 1. Extract context windows (CW) of ±30 terms around each query term from all relevant documents
- 2. Train a LSTM RNN on the characters from the extracted CWs
- 3. Sample characters from the trained LSTM RNN

Output: A synthetic document d

3. Experiments with TREC Disks 4&5, queries 301-450 (only 101 had enough training data).

For each query, ask crowdsourced users to rank wordclouds of three randomly chosen known relevant documents and the RNN-created document.



4. Finding: the wordclouds of documents synthesised by deep learning were assessed by users as overall most relevant (average rank position 1.8).