

Task-Oriented Query Reformulation with Reinforcement Learning

Group 25

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Motivation



Motivation

Query: "deepmind go paper"



[PDF](#) Mastering the game of Go with deep neural networks ... - Go Game G...
<https://gogameguru.com/i/2016/03/deepmind-mastering-go.pdf> ▼
by D Silver - Cited by 756 - Related articles
networks play Go at the level of state-of-the-art Monte-Carlo tree search Ostrovski for reviewing the paper, and the rest of the DeepMind team for their support ...
You visited this page on 3/20/17.



Publications | DeepMind

<https://deepmind.com/publications.html> ▼
Nature 2016. Hybrid computing using a neural network with dynamic external memory. Authors: A Graves, G Wayne, M Reynolds, T Harley, I Danihelka, ...

AlphaGo | DeepMind

<https://deepmind.com/research/alphago/> ▼
Jan 28, 2016 - Featuring expert analysis by Gu Li 9p and Zhou Ruiyang 9p, these games will prove an enlightening read for Go players of all levels.

Mastering the game of Go with deep neural networks and tree search ...

www.nature.com/nature/journal/v529/n7587/full/nature16961.html
by D Silver - 2016 - Cited by 756 - Related articles
Jan 28, 2016 - The game of Go has long been viewed as the most challenging of classic games for artificial intelligence owing to its enormous search space ...

Motivation

Query: "google artificial intelligence paper asian board game"



Master of Go Board Game Is Walloped by Google Computer Program ...

<https://www.nytimes.com/2016/03/10/world/asia/google-alphago-lee-se-dol.html>

Mar 9, 2016 - Lee Se-dol, the world's top player of the boardgame Go, lost the first of five matches to a computer ... Kim Sung-ryong, a South Korean Go master who provided commentary during ... wondered Rodney Brooks, a pioneering artificial intelligence researcher. ... Order Reprints| Today's Paper|Subscribe.

Google AI beats legendary player in Chinese board game - The Hindu

www.thehindu.com › Sci-Tech › Science ▾

Mar 9, 2016 - South Korea's professional Go player Lee Sedol, right, playing the game with against Google's artificial intelligence program, AlphaGo. ... In a new feat, Google-run artificial intelligence (AI) programme "AlphaGo" has defeated legendary player Lee Se-dol in Go — a complex ...

Google AI algorithm masters ancient game of Go : Nature News ...

www.nature.com/news/google-ai-algorithm-masters-ancient-game-of-go-1.19234 ▾

Jan 27, 2016 - Google AI algorithm masters ancient game of Go ... A computer has beaten a human professional for the first time at Go — an ancient board game that ... in Asia, has frustrated the efforts of artificial-intelligence researchers for ...

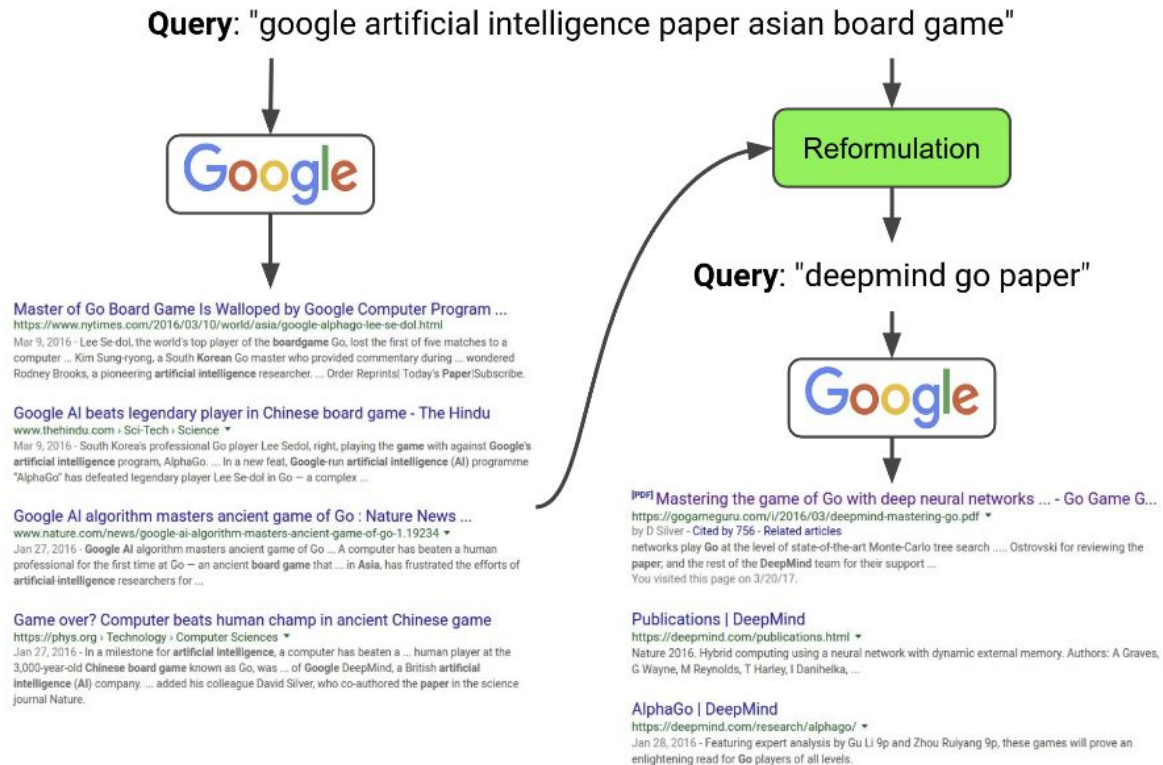
Game over? Computer beats human champ in ancient Chinese game

<https://phys.org> › Technology › Computer Sciences ▾

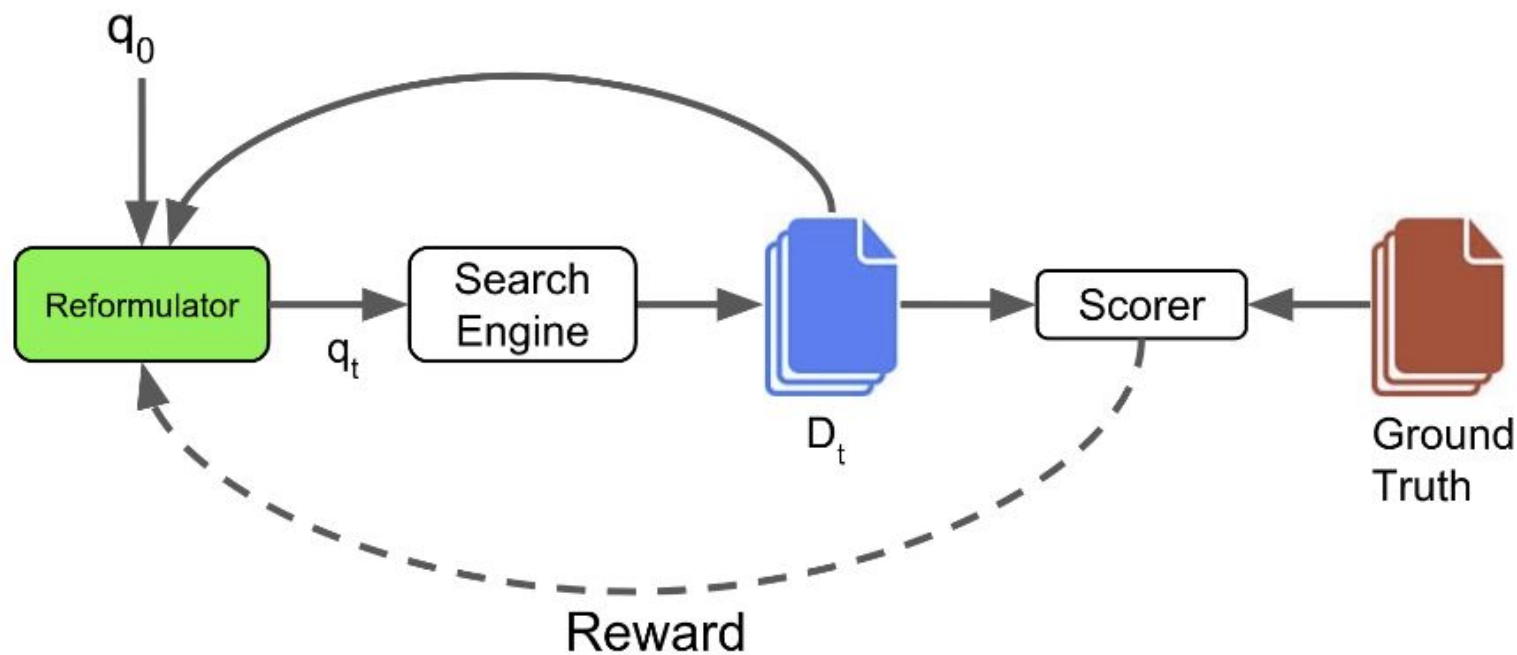
Jan 27, 2016 - In a milestone for artificial intelligence, a computer has beaten a ... human player at the 3,000-year-old Chinese board game known as Go, was ... of Google DeepMind, a British artificial intelligence (AI) company. ... added his colleague David Silver, who co-authored the paper in the science journal Nature.



Idea



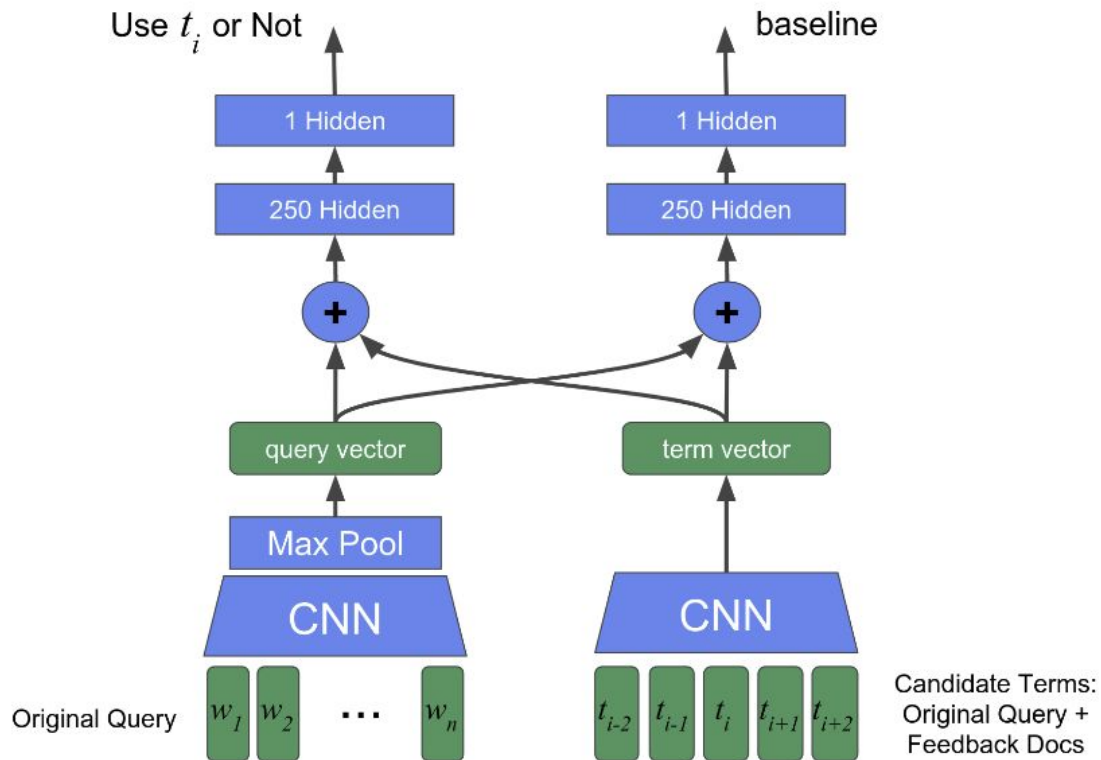
Idea



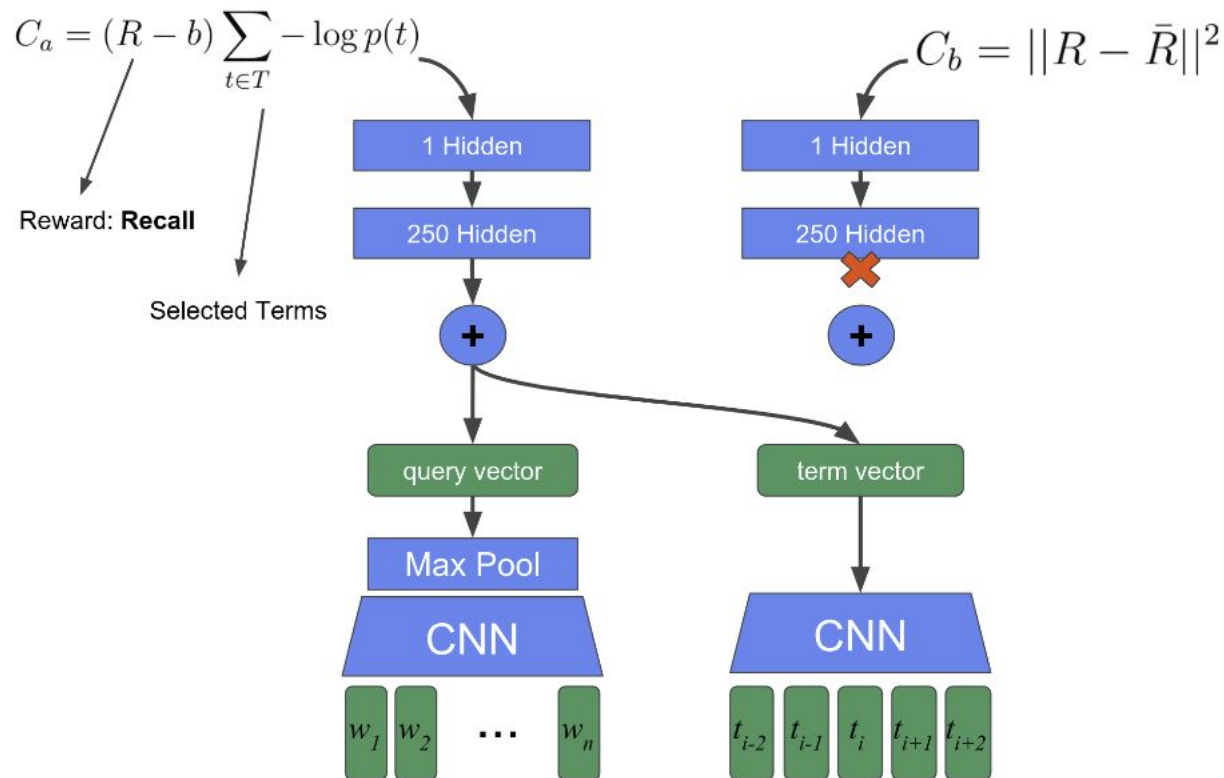
Implementation

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Reformulator



REINFORCE



Oracle - RL

Conservative **Upper-Bound** performance of RL based model:

1. Split validation or test data into smaller subsets (~2000 samples)
2. Overfit RL model on each subset
3. Oracle performance = Average reward over all subsets

Datasets

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TREC - Complex Answer Retrieval Dataset

INPUT: Wikipedia Title + Section
“*Sea Turtle Diet*”

OUTPUT: Wikipedia Paragraphs under the Section

Sea turtle

Diet [edit]

The loggerhead, Kemp's ridley, olive ridley, hawksbill, flatback, and leatherback sea turtles are omnivorous for their entire life. Omnivorous turtles may eat a wide variety of plant and animal life including, [decapods](#), [seagrasses](#), [seaweed](#), [sponges](#), [mollusks](#), [cnidarians](#), [echinoderms](#), [worms](#) and [fish](#).^{[36][37][38][39]} However some species specialize on certain prey.

The diet of green turtles changes with age.^[40] Juveniles are omnivorous, but as they mature they become exclusively herbivorous.^{[37][40]} This diet shift has an effect on the green turtle's morphology.^{[41][42]} Green sea turtles have a serrated jaw that is used to eat sea grass and algae.^[43]

Leatherback turtles feed almost exclusively on jellyfish and help control jellyfish populations.^{[44][45]}

Hawksbills principally eat sponges, which constitute 70–95% of their diets in the Caribbean.^[46]

CORPUS SIZE: 5.9M

Jeopardy Dataset

INPUT: Jeopardy question

“For the last 8 years of his life, Galileo was under house arrest for espousing this man’s theory”

OUTPUT: Wikipedia article whose title is the Answer

Nicolaus Copernicus

From Wikipedia, the free encyclopedia

“Copernicus” redirects here. For other uses, see Copernicus (disambiguation).

Nicolaus Copernicus (/koʊˈpɜːrnɪkəs, kəˈj[[]ˈɪlɪzɪʃ] Polish: Mikołaj Kopernik [miˈkɔwaj koˈpɛrɲik] (help · listen); German: Nikolaus Kopernikus; 19 February 1473 – 24 May 1543) was a Renaissance mathematician and astronomer who formulated a model of the universe that placed the Sun rather than the Earth at the center of the universe, likely independently of Aristarchus of Samos, who had formulated such a model some eighteen centuries earlier.^[a]

The publication of Copernicus' model in his book *De revolutionibus orbium coelestium* (*On the Revolutions of the Celestial Spheres*), just before his death in 1543, was a major event in the history of science, triggering the Copernican Revolution and making an important contribution to the Scientific Revolution.^[7]

Copernicus was born and died in Royal Prussia, a region that had been part of the Kingdom of Poland since 1466. A polyglot and polymath, he obtained a doctorate in canon law and was also a mathematician, astronomer, physician, classics scholar, translator, governor, diplomat, and economist. In 1517 he derived a quantity theory of money – a key concept in economics – and in 1519 he formulated an economics principle that later came to be called Gresham's law.^[8]

Contents [hide]

Nicolaus Copernicus



1580 portrait (artist unknown) in the Old Town City Hall, Toruń

CORPUS SIZE: 3.5M

Microsoft Academic Dataset

Input: Title/Abstract of a Paper

Output: References in that paper

ImageNet Classification with Deep Convolutional Neural Networks

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Abstract

We trained a large, deep convolutional neural network to classify the 1.2 million high-resolution images in the ImageNet ILSVRC-2010 contest into the 1000 different classes. On the test data, we achieved top-1 and top-5 error rates of 37.5% and 17.0%, which is considerably better than the previous state-of-the-art. The neural network, which has 60 million parameters and 650,000 neurons, consists of five convolutional layers, some of which are followed by max-pooling layers, and three fully-connected layers with a final 1000-way softmax. To make training faster, we used non-saturating neurons and a very efficient GPU implementation of the convolution operation. To reduce overfitting in the fully-connected layers we employed a recently-developed regularization method called "dropout" that proved to be very effective. We also entered a variant of this model in the ILSVRC-2012 competition and achieved a winning top-5 test error rate of 15.3%, compared to 26.2% achieved by the second-best entry.

References

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CORPUS SIZE: 500k

Results

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Recall@40

	TREC-CAR	Jeopardy	MS Academic
Original Query	43.6	23.4	12.8
Pseudo Relevance Feedback - TFIDF	44.3	29.9	13.1
Pseudo Relevance Feedback - Word2Vec	44.5	27.5	11.9
Google	-	30.1	-
REINFORCE	47.3	33.4	14.9
Oracle, RL	55.9	42.4	24.6

Samples

Original: *It can be a herdsman's little house in the Swiss Alps, or a ski lodge built in that style*

Reformulated: *house Swiss Alps ski lodge that style castle board
chalet*

Original: Homelessness in Canada, Public Policy

Reformulated: *homelessness in canada public policy human
service programs social policy california treatment of the
homeless numerous*

Discussion

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Discussion

- Oracle shows there is room for improvement
- Applicability: Click-through suggestion for commercial search engines
- Querying the search engine is 90% of training time

References

- Rodrigo Nogueira and Kyunghyun Cho 2017. “***Task-Oriented Query Reformulation with Reinforcement Learning***”
<https://arxiv.org/pdf/1704.04572.pdf>

Questions?
