QUANTUM COMPUTATIONAL LINGUISTICS

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Cambridge Quantum

QUANTINUUM

Solstice of Foundations

ETH Zurich, June 2022



Featuring

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Carys Harvey Douglas Brown Anna Pearson Robin Lorenz

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Vid Kocijan Wenduan Xu

Matty Hoban Dimitri Kartsaklis Stephen Clark Bob Coecke



Computational linguistics and natural language processing

- CL: syntax, grammar, formal language theory, top-down rule-based

- NLP: engineering, big text data, machine learning, bottom-up statistical methods

SotA





GPT-3: 175B params

Language Models are Few-Shot Learners

Tom B. Brown* Benjamin Mann' Nick Ryder* Melanie Subbiah* Jared Kaplan **Prafulla Dhariwal Arvind Neelakantan Pranav Shyam Girish Sastry** Sandhini Agarwal **Gretchen Krueger** Tom Henighan Ariel Herbert-Voss Amanda Askell **Rewon Child** Aditva Ramesh Daniel M. Ziegler Jeffrey Wu **Clemens Winter Christopher Hesse** Mark Chen Mateusz Litwin Eric Sigler Scott Gray **Christopher Berner Benjamin Chess** Jack Clark Sam McCandlish Alec Radford Dario Amodei Ilya Sutskever

OpenAI

LaMDA: 137B params + consulting external knowledge

LaMDA: Language Models for Dialog Applications

Romal Thoppilan Daniel De Freitas * Jamie Hall Noam Shazeer * Apoory Kulshreshtha Heng-Tze Cheng Alicia Jin Taylor Bos Leslie Baker Yu Du YaGuang Li Hongrae Lee Huaixiu Steven Zheng Amin Ghafouri Marcelo Menegali Yanping Huang Maxim Krikun Dmitry Lepikhin James Qin Dehao Chen Yuanzhong Xu Zhifeng Chen Adam Roberts Maarten Bosma Vincent Zhao Yangi Zhou **Chung-Ching Chang** Igor Krivokon Will Rusch Marc Pickett Pranesh Srinivasan Laichee Man Kathleen Meier-Hellstern Meredith Ringel Morris Toju Duke Johnny Soraker Tulsee Doshi **Renelito Delos Santos** Ben Zevenbergen Vinodkumar Prabhakaran Mark Diaz Ben Hutchinson Kristen Olson Alejandra Molina Erin Hoffman-John Josh Lee Lora Aroyo Ravi Rajakumar Alena Butryna Matthew Lamm Viktoriya Kuzmina Aaron Cohen Joe Fenton **Rachel Bernstein** Ray Kurzweil Blaise Aguera-Arcas Claire Cui Marian Croak Ed Chi

Quoc Le

Google

Human Brain: 200 billion neurons, 125 trillion synapses (just in the cerebral cortex). New larger language models with 100 trillion parameters.



a penguin in a pink tshirt chasing a tiger in a black hat



DALL·E mini



`large language models are expert blabbers' – J Bach 'Deep learning models need symbolic structures hardcoded into them' – G Marcus `Deep learning models can eventually learn to do anything only from data' – G Hinton

Quantamagazine Physics Mathematics Biology Computer Science Topics Archive

QUANTIZED COLUMNS

What Does It Mean for AI to Understand?

🤜 60 📔 🥅 It's simple enough for AI to seem to comprehend data, but devising a true test of a machine's knowledge has proved difficult.



Artificial intelligence / Machine learning

OpenAl's new language generator GPT-3 is shockingly good—and completely mindless

The AI is the largest language model ever created and can generate amazing human-like text on demand but won't bring us closer to true intelligence.

by Will Douglas Heaven

July 20, 2020

NOĒMA What AI Can Tell Us **About Intelligence**

Can deep learning systems learn to manipulate symbols? The answers might change our understanding of how intelligence works and what makes humans unique.



"The limits of my language mean the limits of my world" - Ludwig Wittgenstein

"Language faculty is what separates us from other species" - Noam Chomsky

Turing test : verify intelligent behaviour via language!



QUANTUM DISCO THEORY



A hybrid model

Simple, less black-boxy model hopefully provides understandable solution rather than outsource our hope in scaling up.

"What is the mechanics of meaning?"



DisCoCat [1003.4394] Coecke, Sadrzadeh, Clark

Distributional meaning + Compositional structure

'meaning of a word is the words around it'



'grammar can be described by formal rules'





THE BABY DEFINITION OF A CATEGORY

[0905.3010]

A category is:

- a collection of **Objects**.

- with **Arrows** (or 'transformations', or 'morphisms') between them.

Functor: structure preserving map between two cats:

- sends objects to objects and arrows to arrows

- respects composition

Heraclitus-style framework.

It tries to say as little as possible about objects and talks about how they transform. Allows one to think of meaning as analogy.





String diagrams are composed of boxes and wires.

A box represents a process.

Wires are typed.

[Picturing Quantum Processes (dodo book)]

You have to choose a reading direction 'foliation'

process:

objects: types arrows: boxes

output wires t_1 t_2 t_m box * * * wires



Special cases of boxes:

state: in only output effect: in only input scaler: in no input no output

PROCESS THEORIES AND STRING DIAGRAMS

'Cups' and 'caps': special cases of states and effects:



Snake equation: only topology (connectivity) matters:

$$t$$
 t t $wires$
 t $= 1 \Rightarrow can$
 t t $wiggle$



Trace:



Transposition:



PROCESS THEORIES AND STRING DIAGRAMS

Composition: process theories are symmetric monoidal categories

parallel:
$$t_1$$
 t_2 t_1 t_3
parallel: t_2 t_1 t_2 "monoidal" product
t_2 t_4 t_2 t_4

PROCESS THEORIES AND STRING DIAGRAMS

We can **freely compose** as long as we **respect the types**



FUNCTOR = TYPE-WISE AND BOX-WISE SUBSTITUTIONS







Syntactic structures as string diagrams: the parsing of a sentence under a type-based grammar can be viewed as a process and can be drawn as a string diagram.

types : grammar types boxes : type-reduction rules



PREGROUP GRAMMAR

Parser assigns pregrap types to words:

$$\omega \longrightarrow t_{\omega} = t_{i_{1}}^{z_{i_{1}}} \cdot t_{i_{2}}^{z_{i_{2}}} \cdot \cdot t_{i_{|\omega|}}^{z_{i_{|\omega|}}}, \quad z_{i} \in \mathbb{Z} \cong \{\ldots, lll, ll, l, -, r, rr, rr, \ldots\}, \quad t_{i} \in \mathbb{B}$$

Types reduce as:
$$t^{z} \cdot t^{z+1} \rightarrow 1$$
, $1 \rightarrow t^{z+1} \cdot t^{z}$
 $J \qquad J$
 $t^{z} \quad t^{z+1} \quad t^{z}$
 $t^{z} \quad t^{z+1} \quad t^{z}$
 $t^{z+1} \quad t^{z}$

PREGROUP REDUCTIONS AS STRING DIAGRAMS



COMBINATORY CATEGORIAL GRAMMAR

Type reductions:

$$\begin{array}{ccc} \operatorname{FA}(\succ) & \frac{\alpha: X \leftarrow Y & \beta: Y}{\alpha\beta: X} & \operatorname{FC}(B_{\succ}) & \frac{\alpha: X \leftarrow Y & \beta: Y \leftarrow Z}{\alpha\beta: X \leftarrow Z} & \operatorname{FTR}(T_{\succ}) & \frac{\alpha: X}{\alpha: T \leftarrow (X \rightarrowtail T)} \\ & \frac{\alpha: Y & \beta: Y \rightarrowtail X}{\alpha\beta: X} & \operatorname{BC}(B_{\sphericalangle}) & \frac{\alpha: Z \rightarrowtail Y & \beta: Y \rightarrowtail X}{\alpha\beta: Z \rightarrowtail X} & \operatorname{BTR}(T_{\varsigma}) & \frac{\alpha: X}{\alpha: (T \leftarrow X) \rightarrowtail T} \end{array}$$

CG REDUCTIONS AS STRING DIAGRAMS





Semantics-respecting diagram rewrites



SEMANTIC FUNCTOR





Book: Picturing Quantum Processes Coecke and Kissinger

Quantum theory is a process theory too!

Types : qubit Hilbert spaces (complex vector spaces of dim $2^n, n \in \mathbb{N}$) Boxes : Spiders (linear maps) [0906.4725]: foundational work by Coecke and Duncan

Z-spider:

$$V = 10...0 \times 0...01 + e^{i\alpha} 1...1 \times 1...1 , \quad \alpha \in \mathbb{E}_{0,2\pi}$$

$$V = 10...0 \times 0...01 + e^{i\alpha} 1....1 \times 1...1 , \quad \alpha \in \mathbb{E}_{0,2\pi}$$

$$U = 10...0 \times 0...01 + e^{i\alpha} 1....1 \times 1...1 , \quad \alpha \in \mathbb{E}_{0,2\pi}$$



[2012.13966]

A.5 ZX-calculus full cheatsheet

The following rewrite rules hold for all $\alpha, \beta, \alpha_i, \beta_j, \gamma_k \in \mathbb{R}$ and $a \in \{0, 1\}$ (up to global non-zero scalar).



QUANTUM PROCESSES

е×

The standard Clifford+T states and gates are built out of spiders.

m = 10 m = 11 0 = 1+ n = 1-





Quantum states act as word embeddings.

,



Ju; trained from data



$$\frac{1}{2} \left[\left(\frac{1}{2} \omega_{i} \right) \right] = \left[\left(\frac{1}{2} \omega_{i} \right) \left(\frac{1}{2} \omega_{i} \right) \right]^{2} = \left[\left(\frac{1}{2} \omega_{i} \right) \left(\frac{1}{2} \omega_{i} \right) \right]^{2} = \left[\left(\frac{1}{2} \omega_{i} \right) \left(\frac{1}{2} \omega_{i} \right) \right]^{2} = \left[\left(\frac{1}{2} \omega_{i} \right) \left(\frac{1}{2} \omega_{i} \right) \right]^{2} = \left[\left(\frac{1}{2} \omega_{i} \right) \left(\frac{1}{2} \omega_{i} \right) \left(\frac{1}{2} \omega_{i} \right) \right]^{2} = \left[\left(\frac{1}{2} \omega_{i} \right) \left(\frac{1}{2} \omega_{i} \right) \left(\frac{1}{2} \omega_{i} \right) \right]^{2} = \left[\left(\frac{1}{2} \omega_{i} \right) \left(\frac{1}{2} \omega_$$

These word-similarities can go into a glove cost func and θ_{w_i} are trained on big text-data from which a cooc matrix X_{ij} is obtained.

$$\min_{\theta} \left| \left\langle 0 \left| U(\theta_i) U^{\dagger}(\theta_j) \right| 0 \right\rangle \right|^2 + b_i + b'_j - \log X_{ij} \qquad \theta = \bigcup_i \theta_i$$



We can estimate $|\langle 0|U(\theta_{w_i})U^{\dagger}(\theta_{w_j})|0\rangle|^2$ up to additive error with $O(|V|^2)$ QC calls.





Type reductions are mapped to quantum processes.

pregroups

$$t \quad t' \quad t \Rightarrow 9_{t} \quad z' \quad y' \quad t'' \quad = \quad \bigcup \quad = \quad \bigcup \quad Bell \quad effect$$

 \bigcirc = \bigcirc = |

QUANTUM LINGUISTIC PROCESSESLambeq [2110.04236]

 $ex: 9t^{=1}, \forall t$



QUANTUM LINGUISTIC PROCESSES

Type reductions are mapped to quantum processes.

CCG



Ja: trained in an NLP task

QUANTUM LINGUISTIC PROCESSESLambeq [2110.04236]

 $ex: q_t = 1, \forall t$









EXPERIMENTS IN THE QUANTUM DISCO



[2005.02975] DisCoPy: Monoidal Categories in Python

Giovanni de Felice, Alexis Toumi, Bob Coecke Department of Computer Science, University of Oxford. Cambridge Quantum Computing Ltd. {firstname.lastname}@cs.ox.ac.uk

We introduce DisCoPy, an open source toolbox for computing with monoidal categories. The library provides an intuitive syntax for defining string diagrams and monoidal functors. Its modularity allows the efficient implementation of computational experiments in the various applications of category theory where diagrams have become a lingua franca. As an example, we used DisCoPy to perform natural language processing on quantum hardware for the first time.

[2205.05190]

DisCoPy for the quantum computer scientist

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† Cambridge Quantum Computing Ltd.
* Department of Computer Science, University of Oxford

May 12, 2022

Abstract

DisCoPy (Distributional Compositional Python) is an open source toolbox for computing with string diagrams and functors. In particular, the diagram data structure allows to encode various kinds of quantum processes, with functors for classical simulation and optimisation, as well as compilation and evaluation on quantum hardware. This includes the ZX calculus and its many variants, the parameterised circuits used in quantum machine learning, but also linear optical quantum computing. We review the recent developments of the library in this direction, making DisCoPy a toolbox for the quantum computer scientist.

https://github.com/oxford-quantum-group/discopy pip install discopy

[2110.04236]

https://github.com/CQCL/lambeq pip install lambeq

lambeq: An Efficient High-Level Python Library for Quantum NLP

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Abstract

We present lambeq, the first high-level Python library for Quantum Natural Language Processing (QNLP). The toolkit offers a detailed hierarchy of modules and classes implementing all stages of a pipeline for converting sentences to string diagrams, tensor networks, and quantum circuits ready to be used on a quantum computer. lambeq supports syntactical parsing, rewriting and simplification of string diagrams, ansatz creation and manipulation, as well as a number of compositional models for preparing quantum-friendly representations of sentences, employing various degrees of syntax sensitivity. We present the generic architecture and we describe the most important modules in detail, demonstrating the usage with illustrative examples. Further, we test the toolkit in practice by using it in order to perform a number of experiments on simple NLP tasks, implementing both classical and quantum pipelines, with success.

Online demo: https://qnlp.cambridgequantum.com/generate.html



λambeq enables the automatic deployment of large-scale compositional language models



[2105.07720]



Supervised learning





.



TOY DATA: CLASSICAL SIM

[2012.03756]

Pregroup model

|V| = 7

15 train sents 15 test sents

('Dude who loves Walter bowls', 1), ('Dude bowls', 1), ('Dude annoys Walter', 0), ('Walter who abides bowls', 0), ('Walter loves Walter', 1), ('Walter annoys Dude', 1), ('Walter abides', 0), ('Dude loves Walter', 1), ('Dude who bowls abides', 1), ('Walter who bowls annoys Dude', 1), ('Dude who bowls bowls', 1), ('Dude who abides abides', 1), ('Dude annoys Dude who bowls', 0), ...

Note:

Intra-sentence correlations are 'quantum': due to grammar Inter-sentence correlations are 'classical': due to shared words



IQP ansatz





[2012.03756]

Pregroup model

 $q_n = 1$, $q_s = 0$, IQP ansatz

|V|= 5

8 train sents 8 test sents

False:

(Juliet kills Romeo who dies, 0) (Romeo kills Juliet, 0) (Romeo who kills Juliet dies, 0)

•••

•••

True: (Juliet dies, 1) (Romeo who dies loves Juliet, 1) (Romeo who kills Romeo dies, 1)



Ibmq_montreal

$$\lfloor (\theta) = \leq \left(l_s^{pr}(\theta) - l_s^{data} \right)^2$$

TOY DATA: CLASSICAL SIM

person

Pregroup model

 $q_n = 1, q_s = 1$, IQP Ansatz, d = 1

Train sents: 70 Dev sents: 30 Test sents: 30

Cooking: (Skilful man prepares sauce, 0) (Woman cooks tasty meal, 0) (Skilful person prepares meal, 0) ...

Technology: (Skilful woman debugs program, 1) (Man prepares useful application, 1) (Person debugs useful software, 1)

•••



dinner

1tasty

prepares



$$C_{ost}(\theta) = \leq l_s \log l_s^{pred} + (|-l_s^{(\theta)})\log (|-l_s^{pred})$$

TOY DATA: QUANTUM RUN



Same topic classification task.

Train classically, not even considering noise.

Then Test on H1 (12 qbts).

Acc 97%.

Ibmq_bogota



Lambeq'stree-reader [2110.04236]







Unibox





Rule-based

QUANTUM TREE SPECIES





" Syntactic QCNN " [1810.03787]

> no barren platea us for tree gairas [2011.06258]



TOY DATA: CLASSICAL SIM





REAL-WORLD DATA: CLASSICAL SIM

Tree species: unibox t : 3qb ∀t ansatz: 2-layer IQP dataset: rt-polarity Train: 216 sents Test: 35 sents

word states trained in-task

"A masterful film from a master filmmaker, unique in its deceptive grimness, compelling in its fatalist worldview."

"if you love reading and/or poetry, then by all means check it out.you'll probably love it."

"... stumbles over every cheap trick in the book trying to make the outrage come even easier "

"the only way to tolerate this insipid , brutally clueless film might be with a large dose of painkillers " Validation accuracy: 0.5714285714285714





[1904.03478]

Quantum DisCoCirc: sentence composition; satisfies specific desiderata of compositionality.

Bob Coecke: "get quadvantage a la quantum simulation". So **IF** this model works well for some tasks, one would need a QC to run it.











Sandra likes Mary. Sandra journeyed to the garden. Mary went to the bathroom. Mary had a shower







C; Overlaps of higher-oder processes

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\end{array}$







G



similarity between nouns NIN2 after the modifiers acted on them



ģ

text similarity



or do swap test

 $p_{1} - p_{1} - p_{1$ additive emer Endes $\left\| P(\circ) - |\langle T_1 | T_2 \rangle|^2 \right\| \leq \varepsilon$

if some nouns are different between TI, TZ, You can't compare the fexts. But maybe if they share Some norm, we can discard the ones that are not common





 $if \quad C_{T_i}^{(M)} C_{T_2}^{(M)} \sim I$ then texts are the some process on any input hours QMA - hard (non-I check) (exp-many samples wort ase)

DQC1 for trace estimation



Join the QNLP Discord server!

https://discord.gg/zW9zHNpdnN

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quantinuum.com/careers