Epilogue

What did we cover?

- Language Modeling:
 - Markov language models, neural language models: feed-forward, RNN, Transformers
- Classification:
 - The basics: discriminative (log-linear) and generative models (naive Bayes)
 - Sequence labeling: HMM, MEMM, CRF for POS tagging and NER mostly
- Text Classification and Sentiment Analysis
- Grammar
 - What is grammar?
 - Syntactic Parsing: graph-based, transitionbased

- Broad-coverage semantic analysis:
 - Information Extraction
 - Indirect supervision
 - Semantic role labeling
 - Syntax and semantics
- Machine Translation (MT)
 - Nature of the task
 - Neural MT
- Contextualized word embeddings
 - Attention and self-attention
 - The Transformer model
 - Pretraining
 - Masked language models

Many More Topics to Cover

• Semantics:

- Lexical Semantics
- Sentence semantics: we touched on this, but much more to cover
- Executable semantic forms: text to code, personal assistants (connection to apps and APIs)
- Discourse Structure: the relation between sentences
- Question answering

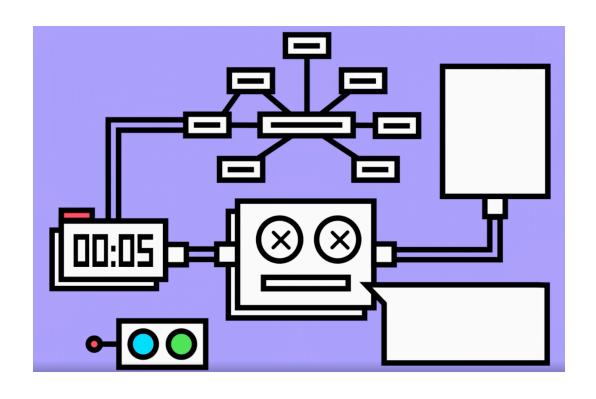
```
Complex
question
What are the name and budget of the departments
with average instructor salary greater than the
overall average?

Complex
SQL
SELECT T2.name, T2.budget
FROM instructor as T1 JOIN department as
T2 ON T1.department_id = T2.id
GROUP BY T1.department_id
HAVING avg(T1.salary) >
(SELECT avg(salary) FROM instructor)
```

Many More Topics to Cover

- Dialogue systems: one of the most important frontiers of the field, both theoretically and practically
 - Chit-chat
 - Task-oriented dialogue: sales agents, customer support
 - Assistance
 - Debate





https://www.theverge.com/2018/6/13/17453994/amazon-

https://www.nature.com/articles/s41586-021-03215-w

Many More Topics to Cover

- Deep learning for NLP
- Resources
- Evaluation
- Cognitive modeling: psycholinguistics, computational neuroscience

• ...

NLP: Some Big Questions

- Universal semantic metalanguage
- How do languages differ from one another?
- Multi-modal systems (e.g., visual, auditory, tactile)
- Implicit Information (implicit claims, implicit participants, sub-text)
- Subjective text (answering non-factual questions, understanding nuances, generating them, aesthetic aspects of language)

What is the Role of Linguistic Structure in Present Day NLP?

- Say what you will about neural language models, their output is mostly fluent
- What happens if we take a closer look?
 - Slice it by different syntactic phenomena
 - This is very much an open question

The BLiMP Benchmark

Phenomenon	UID	5-gra	m LST	M TXI	GPT	2 Hun	Acceptable Example	Unacceptable Example
ANAPHOR	anaphor_gender_agreement anaphor_number_agreement	44	88	91	99	96	Katherine can't help herself . Many teenagers were helping themselves .	Katherine can't help himself.
AGREEMENT	anaphor_number_agreement	52	95	97	100	99	Many teenagers were helping themselves.	Many teenagers were helping herself.
STRUCTURE	animate_subject_passive	54	68	58	77	98	Amanda was respected by some waitresses. Danielle visited Irene.	Amanda was respected by some picture.
	animate_subject_trans	72	79	70	80	87	Danielle visited Irene.	Amanda was respected by some picture . The eye visited Irene.
	causative	51	65	54	68	82	Aaron breaks the glass. The Lutherans couldn't skate around.	Aaron appeared the glass. The Lutherans couldn't disagree with.
	drop_argument	68	79	67	84	90		The Lutherans couldn't disagree with.
	inchoative	89	72	81	90	95	A screen was fading. Some glaciers are vaporizing.	A screen was cleaning. Some glaciers are scaring.
	intransitive	82	73	81	90	86	Some glaciers are vaporizing.	Some glaciers are scaring.
	passive_1	71	65	76	89	99	Some glaciers are vaporizing. Jeffrey's sons are insulted by Tina's supervisor. Most cashiers are disliked.	Jeffrey's sons are smiled by Tina's supervisor. Most cashiers are flirted.
	passive_2	70	72	74	79	86	Most cashiers are disliked.	Most cashiers are flirted.
	transitive	91	87	89	49	87	A lot of actresses' nieces have toured that art gallery.	A lot of actresses' nieces have coped that art gallery.
BINDING	principle_A_c_command	58	59	61	100	86	A lot of actresses that thought about Alice healed themselves.	A lot of actresses that thought about Alice healed herself.
	principle_A_case_1	100	100	100	96	98	A lot of actresses that thought about Alice healed themselves . Tara thinks that she sounded like Wayne.	Tara thinks that herself sounded like Wayne.
	principle_A_case_2	49	87	95	73	96	Stacy imagines herself praising this actress.	Stacy imagines herself praises this actress. Carlos said that Lori helped himself .
	principle_A_domain_1	95	98	99	99	95	Carlos said that Lori helped him.	Carlos said that Lori helped himself.
		56	68	70	73	75	Mark imagines Erin might admire herself . Nancy could say every guy hides himself .	Mark imagines Erin might admire himself. Every guy could say Nancy hides himself.
	principle_A_domain_2 principle_A_domain_3 principle_A_reconstruction	52	55	60	82	83	Nancy could say every guy hides himself.	Every guy could say Nancy hides himself.
	principle_A_reconstruction	40	46	38	37	78	It's herself who Karen criticized.	It's herself who criticized Karen.

Do Neural Language Models Implicitly Learn Grammar?

- Current answer: depends on how big your model is and who you ask
- Can we just read the representation that these networks ascribe to sentence from the network activations?
 - How?
 - Does grammar have a role in the next phase of NLP?
- Big questions arise...
 - What part of grammar do neural LMs capture?
 - Do they have systematic biases?
 - Are distributional signals sufficient to learn grammar?
 - Do findings generalize cross-linguistically?
 - Does interpretability matter?

Courses

- In NLP:
 - Advanced course in NLP
 - Seminars by Roy Schwartz, Gabi Stanovsky and myself
 - Advanced students are welcome to join our reading group
- Other related courses:
 - Introduction to Information Theory
 - Machine learning courses
 - Probabilistic methods in artificial intelligence
 - Deep Learning, Cognition and Brain (a course in Cog-Sci)