# Accelerating Science: A Grand Challenge for Al

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Allen Institute for Artificial Intelligence

18 November, 2016



# Allen Institute for AI (AI2)



#### Founded by Paul Allen in 2014 Now 70 people and growing



#### Al for the Common Good



### Moore's Law of Scientific Publication

# The number of scientific papers has doubled every nine years since World War II





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Google	deep learning	
Scholar	About 3,810,000 results 0.04 sec)	
Articles Case law My library	<b>Learning in science:</b> A comparison of <b>deep</b> and surface approaches C Chin, DE Brown - Journal of research in science teaching, 2000 - Wiley Online Library Abstract The purpose of this study was to explore in greater depth what has been called by previous researchers, a <b>deep</b> versus surface approach to <b>learning</b> science. Six Grade 8 students judged as typically using <b>learning</b> approaches ranging from <b>deep</b> to surface were Cited by 397 Related articles All 5 versions Cite Save	
Any time Since 2016 Since 2015 Since 2012 Custom range	Why does unsupervised pre-training help deep learning? <u>D Erhan, Y Bengio, A Courville</u> , PA Manzagol of Machine Learning, 2010 - jmlr.org Abstract Much recent research has been devoted to learning algorithms for deep architectures such as Deep Belief Networks and stacks of auto-encoder variants, with impressive results obtained in several areas, mostly on vision and language data sets Cited by 722 Related articles All 27 versions Cite Save	
Sort by relevance Sort by date	[PDF] Multimodal deep learning J Ngiam, <u>A Khosla</u> , M Kim, <u>J Nam</u> machine learning (, 2011 - machinelearning.wustl.edu Abstract Deep networks have been successfully applied to unsupervised feature learning for single modalities (eq. text. images or audio). In this work, we propose a powel application of	
<ul> <li>✓ include patents</li> <li>✓ include citations</li> </ul>	deep networks to learn features over multiple modalities. We propose a nover application of Cited by 519 Related articles All 29 versions Cite Save More	



### Information Overload



# **Challenge**: Researchers are swamped; Virtually impossible to read all papers

#### **Opportunity:** Leverage AI to combat information overload



### Semantic Scholar (S2)



Home in on key papers, citations, and results.

Q Find it fast

Try: Open information extraction POS tagging Dependency parsing

Computer Science and (recently) Neuroscience research articles



### Semantic Scholar

#### <u>Demo</u>



### Faceted Search

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### Faceted Search

Semantic Scholar		Songbird Basal Ganglia	📑 SIGN IN  🥹	
Filter Results:	_	Page 1	Sort by: Relevance	
Field of Study	~	Incomplete and Inaccurate Vocal Imitation after Knockdown of FoxP2 in	in Songbird Basal	
Publication Year	~	Ganglia Nucleus Area X Sebestian Macelar, Christella Bashafart, Basiamin Caarri, Brund Liamamid, Bauel Onton, Caratanae Sebarff, Bl aS biology, 2007		
Publication Type	~	The gene encoding the forkhead box transcription factor, FOXP2, is essential for developing the full an language. Mutations of FOXP2 cause developmental verbal dyspraxia (DVD), a speech and language (	rticulatory power of human	
Author	~	fluent production of words and the correct use and comprehension of grammar. FOXP2 patients have.	(More)	
Key Phrase	~			
Publication Venue	~	Organization of the songbird basal ganglia, including area X.		
Brain Region	~	Abigail L Person, Samuel D Gale, Michael A Farries, David J Perkel • The Journal of comparative neuro Area X is a songbird basal ganglia nucleus that is required for vocal learning. Both Area X and its imm	blogy - 2008 mediate surround, the medial	
Cell Type	~	striatum (MSt), contain cells displaying either striatal or pallidal characteristics. We used pathway-traci the targets of Area X and MSt with those of the lateral striatum (LSt) and globus (More)	ing techniques to compare directly	
Method	~	• 1 C View On PubMed Related Publications More		
Organism		Automatically extracted using statistical m	nodels	
U Zebra Finch (616)		Vocal learning in songbirds requires a <b>basal ganglia</b> circuit termed the anterior forebrain pathway (AFF song production, and its role in song learning is not well understood. Like the mammalian striatum, the Area X, receives dense dopaminergic innervation from the midbrain. Since dopamine (DA) (More)	P). The AFP is not required for e striatal component of the AFP,	
Rat (24)		View PDF Related Publications More		
Mouse (14)		Vocal Experimentation in the Juvenile Senabird Requires a Recal Con-	alia Circuit	
Drosophila (5)		Bence P. Ölyeczlov, Aaron S. Andelman, Michale S. Ees, Pl. oS. biology, 2005	gila Olicult	
C. elegans (2)		Songbirds learn their songs by trial-and-error experimentation, producing highly variable vocal output i own sounds to the song of a tutor, young songbirds gradually converge to a stable song that can be a tutor song. Here we show that vocal variability in the learning <b>songbird</b> is induced by a (More)	as juveniles. By comparing their a remarkably good copy of the	



### Extracted Key Elements





# Staying up-to-date



Cornell University Library

arXiv.org > cs

#### **Computer Science**

Authors and titles for recent submissions, skipping first 205

- Thu, 17 Nov 2016
- Wed, 16 Nov 2016
- Tue, 15 Nov 2016
- Mon, 14 Nov 2016
- Fri, 11 Nov 2016

[ total of 516 entries: 1–25 | ... | 131–155 | 156–180 | 181–205 | **206–230** | 231–255 | 256–280 | 281–305 | ... | 506–516 ] [ showing 25 entries per page: fewer | more | all ]

Tue, 15 Nov 2016 (showing first 25 of 158 entries)

[206] arXiv:1611.04581 [pdf, other]

How to scale distributed deep learning? Peter H. Jin, Qiaochu Yuan, Forrest landola, Kurt Keutzer Comments: Extended version of paper accepted at ML Sys 2016 (at NIPS 2016) Subjects: Learning (cs.LG)



- Idea: Daily feed of most relevant papers
- Research Challenges:
  - Identify topics in a documents
  - Model user's topic preference
  - Rank by relevance



# Deeper Understanding of a Document

autoencod	lers
<u>-</u>	
Induction	
	tructure
-	upervised st

d prediction

**CoNLL 2007** 

Entity Extraction



# Deeper Understanding of a Document



Type Identification



# Deeper Understanding of a Document



#### **Relation Extraction**



# Deeper Understanding of the Literature



#### **Literature Graph**



## Citeomatic: Who should I cite?

- Disseminating Research by Writing papers
  - Comprehensive review of related work is challenging



# Citeomatic: A brief overview

- Input:
  - The title and abstract of a query paper
- Output:
  - A list of related work, that should be reviewed, ordered by confidence



### Citeomatic





### Citeomatic: Network Architecture





Feature / Idea	Addresses Information Overload in	<b>Research Challenges</b>
Faceted Search	Finding Relevant Documents	Search
Automatically extracted important elements	Reading Relevant Documents	Information Extraction (IE)
Daily Feed	Staying Up-to-date	Recommendation
Literature Graph	Understanding and Relating to existing Knowledge	NER, KB Construction etc.
Citeomatic	Writing and Publishing	Document Similarity



# Directions of Future Work

- Hypothesis Generation
  - Method X has been found to be effective for task Y. Task Y and Z are related. Maybe X can be applied to Z
- What's at my knowledge-frontier?
  - I know about X and Y, what should I know next?
- Topic dependencies
  - I want to learn X, what are its pre-requisites?



## Conclusion

"No human could possibly read the entirety of medical literature, personal health records, and case file histories that might inform a doctor's professional opinion when trying to save a cancer patient's life. But a machine can."

- fortune.com on Nov 2, 2016

Scientific Research is facing tremendous Information Overload

Advances in AI, ML and NLP can help !



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