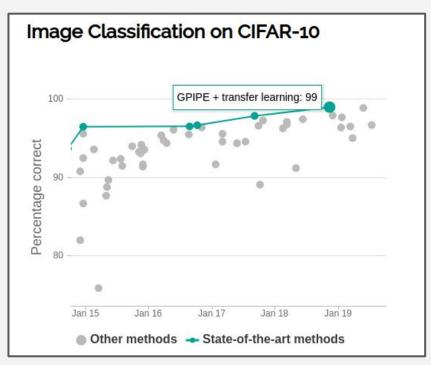
# Al for Common Good through benchmarking at scale

Laure Delisle



# Al research is doing great!

#### Performant



	Comparison to best public-available rest				
	Model	Acc.	#Param	Our Model	Acc.
CIFAR-10	NASNet-A	98.0%	85M	EfficientNet-B0	98.1%
CIFAR-100	NASNet-A	87.5%	85M	EfficientNet-B0	88.1%
Birdsnap	Inception-v4	81.8%	41M	EfficientNet-B5	82.0%
Stanford Cars	Inception-v4	93.4%	41M	EfficientNet-B3	93.6%
Flowers	Inception-v4	98.5%	41M	EfficientNet-B5	98.5%
FGVC Aircraft	Inception-v4	90.9%	41M	EfficientNet-B3	90.7%
Oxford-IIIT Pets	ResNet-152	94.5%	58M	EfficientNet-B4	94.8%
Food-101	Inception-v4	90.8%	41M	EfficientNet-B4	91.5%

Benchmarking

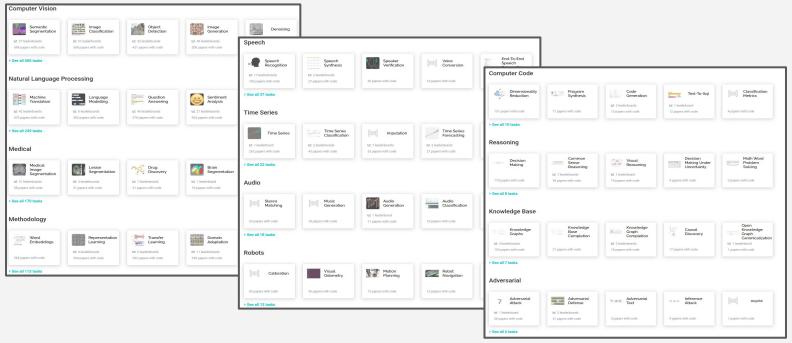
Performance

**SOTA** - Papers with Code

#### Prolific

#### Browse state-of-the-art

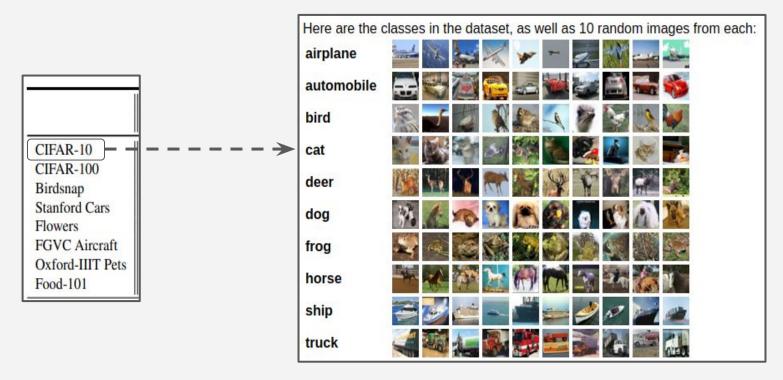
1407 leaderboards • 1291 tasks • 1282 datasets • 16093 papers with code



State Of The Art

#### We can do even better

#### Current benchmarking datasets...



#### Datasets for real world impact

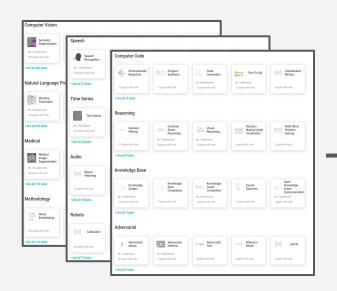




#### Underserved datasets

Cassava leaves dataset - Ramcharan et al. <u>Deep Learning for Image-Based Cassava Disease Detection</u>. Frontiers in Plant Science 2017. Ethiopian datasets - <u>Knoema</u>. Illustration: <u>Ethiopian anthem</u> in Ahmaric Somali audio transcription dataset - Shagodoon. 2019

#### Filling the gap

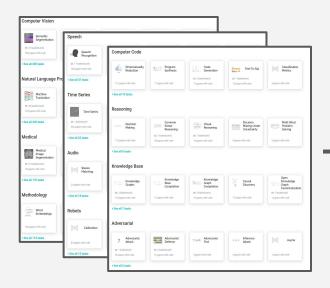






Underserved datasets

#### Filling the gap



- Incentives ➤



State Of The Art

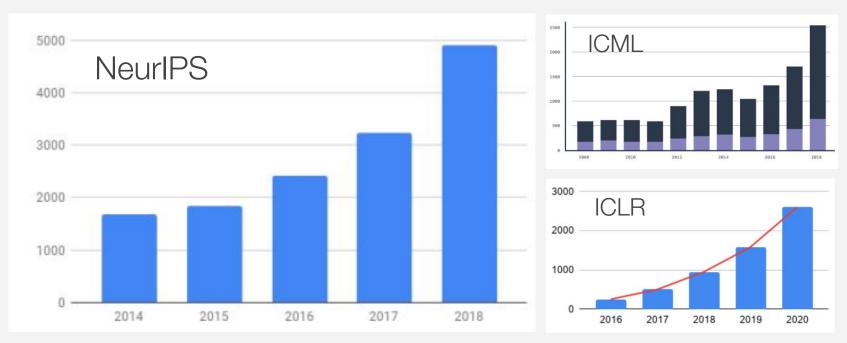
Underserved datasets

## Exploiting the system

In 2018, the number of conference paper submissions has increased

by 47% for ICML, by 50% for NeurIPS, and by almost 100% for ICLR.

#### Al papers submitted

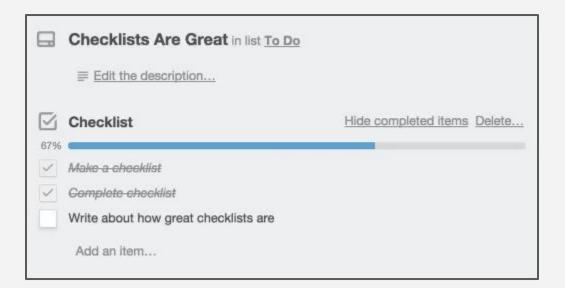


**NeurIPS** - D. Sculley, Jasper Snoek and Alex Wiltschko. <u>Avoiding a Tragedy of the Commons in the Peer Review Process</u>. Critiquing and Correcting Trends in Machine Learning, NeurIPS 2018 Workshop.

ICML - Peltarion. <a href="https://peltarion.com/article/icml-2018-an-ai-party-in-our-own-backyard">https://peltarion.com/article/icml-2018-an-ai-party-in-our-own-backyard</a>

ICLR - @iclr\_conf. Tweet on Sep 25 2019.

#### Checking boxes



Checklist illustration - <u>Trello</u>
NeurlPS checklist - @blaine\_bateman. <u>Tweet</u> on May 9 2019.



#### The Machine Learning Reproducibility Checklist (required for all NeurIPS 2019 papers) buff.ly/2Y8tYnP

	e Machine Learning Reproducibility Checklist (Version 1.2, Mar.27 2019
For	all models and algorithms presented, check if you include:
	A clear description of the mathematical setting, algorithm, and/or model.
	An analysis of the complexity (time, space, sample size) of any algorithm.
	A link to a downloadable source code, with specification of all dependencies, including
	external libraries.
For	any theoretical claim, check if you include:
	A statement of the result.
	A clear explanation of any assumptions.
	A complete proof of the claim.
For	all <b>figures</b> and <b>tables</b> that present empirical results, check if you include:
	A complete description of the data collection process, including sample size.
	A link to a downloadable version of the dataset or simulation environment.
	An explanation of any data that were excluded, description of any pre-processing step.
	An explanation of how samples were allocated for training / validation / testing.
	The range of hyper-parameters considered, method to select the best hyper-parameter
	configuration, and specification of all hyper-parameters used to generate results.
	The exact number of evaluation runs.
	A description of how experiments were run.
	A clear definition of the specific measure or statistics used to report results.
	Clearly defined error bars.
	A description of results with <u>central tendency</u> (e.g. mean) & <u>variation</u> (e.g. stddev).
	A description of the computing infrastructure used.

Reproduced from: www.cs.mcgill.ca/~jpineau/ReproducibilityChecklist.pdf

### ImpactNet

#### ImpactNet by (AI) COMMONS

