

# NeurIPS Summary

CMS MLJC, 10th Dec 2018

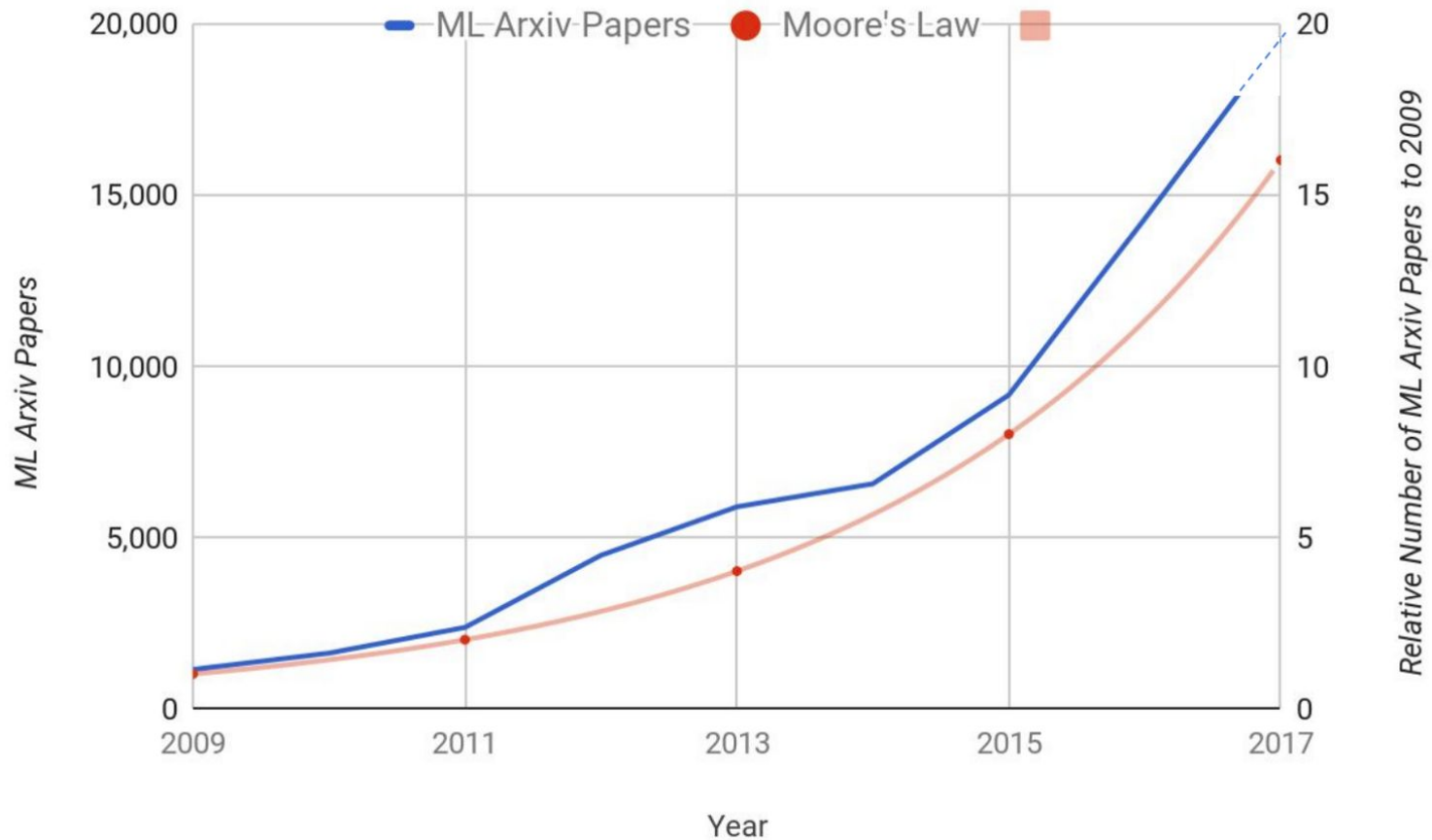
# General Comments

- In Montreal, Canada
- Officially changed name NIPS -> NeurIPS
- Number of Submissions: > 4000
- ~ 800 accepted
- Sold out in < 12 minutes

# Plenary Talks: Highlights

- Unsupervised Learning by Alex Graves:  
<https://www.youtube.com/watch?v=rjZCjosEFpI>
- Visualization in ML: UMAP:
  - Faster than TSNE
  - Seems to capture global structure better
  - Clusters are pulled away more efficiently

# ML Arxiv Papers per Year



# Best contributions according to reviewers

- Non-Delusional Q-learning and Value Iteration - Lu et. al.
- Optimal Algorithms for Non-Smooth Distributed Optimization in Networks - Scaman et.al.
- Nearly Tight Sample Complexity Bounds for Learning Figures of Gaussians via Sample Compression Schemes - Ashtiani et. al.
- **Neural Ordinary Differential Equations - Chen et. al.**

# General Contributions to Bookmark

- “Recent Advances in Autoencoder-Based Representation Learning”:  
<https://www.nari.ee.ethz.ch/commth/pubs/files/autoenc2018.pdf>
- “Understanding Batch Normalization”: <https://arxiv.org/abs/1806.02375>
- “The Effect of Network Width on the Performance of Large-batch Training”:  
<https://arxiv.org/abs/1806.03791>
- “How to Start Training: The Effect of Initialization and Architecture”: <https://arxiv.org/abs/1803.01719>
- “L4: Practical loss-based stepsize adaptation for deep learning”: <https://arxiv.org/abs/1802.05074>
- ...

## Isolation Sources of Disentanglement of VAEs

- Disentangled representation study

$$\mathcal{L}_\beta = \frac{1}{N} \sum_{n=1}^N (\mathbb{E}_q[\log p(x_n|z)] - \beta \text{KL}(q(z|x_n)||p(z)))$$

## Isolation Sources of Disentanglement of VAEs

- ELBO decomposed to show magnify loss on total correlation between latent variables

$$\mathbb{E}_{p(n)} \left[ \text{KL}(q(z|n) || p(z)) \right] = \underbrace{\text{KL}(q(z, n) || q(z)p(n))}_{\text{(i) Index-Code MI}} + \underbrace{\text{KL}(q(z) || \prod_j q(z_j))}_{\text{(ii) Total Correlation}} + \underbrace{\sum_j \text{KL}(q(z_j) || p(z_j))}_{\text{(iii) Dimension-wise KL}} \quad (2)$$

$$\mathcal{L}_{\beta\text{-TC}} := \mathbb{E}_{q(z|n)p(n)} [\log p(n|z)] - \alpha I_q(z; n) - \beta \text{KL}(q(z) || \prod_j q(z_j)) - \gamma \sum_j \text{KL}(q(z_j) || p(z_j)) \quad (4)$$