



ICML @ NYC

International Conference on Machine Learning

JUNE 19-24 2016 NEW YORK

CONFERENCE REPORT

MARCIN ZADROGA



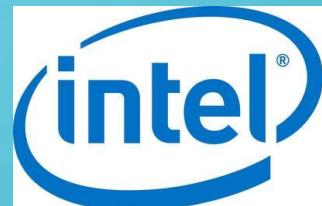
CONFERENCE

- 33rd edition
- 19–24 June 2016 (6 days, sic!)
- Marriot Marquis Hotel on Time Square
- 3000 participants, 1 day of tutorials (9), 3 days of conference (300+ accepted papers), 2 days of workshops (23)

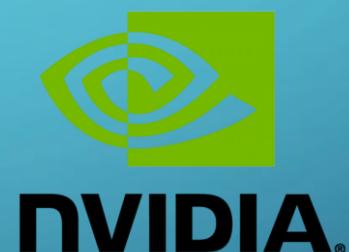
BIG TOPICS ON THE CONFERENCE

- Neural Networks & Deep Learning
- (Deep) Reinforcement Learning
- Speech Recognition
 - (Google) “Towards End-To-End Speech Recognition with Recurrent Neural Networks”
 - (Baidu) “Deep Speech 2: End-to-End Speech Recognition in English and Mandarin”
- Alpha GO
- Machine Learning Systems and Old Ideas revision
- Other: Kernel Methods, Gaussian Processes, Time-Series Analysis, Large Scale Learning

(SOME) COMPANIES PRESENT AT ICML



NETFLIX



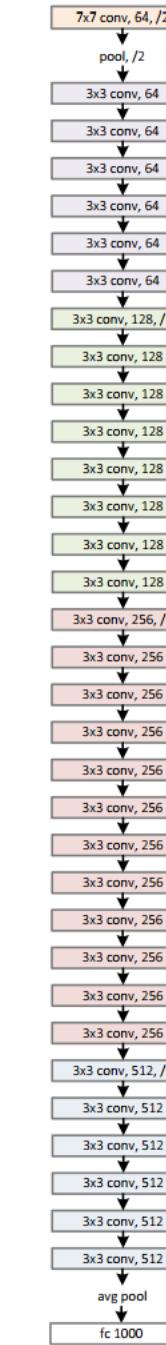
TUTORIALS – HIGHLIGHTS

- Deep Learning
 - <https://github.com/KaimingHe/deep-residual-networks>
 - http://icml.cc/2016/tutorials/icml2016_tutorial_deep_residual_networks_kaiminghe.pdf
- Reinforcement Learning
 - ALPHA GO
 - http://icml.cc/2016/tutorials/deep_rl_tutorial.pdf

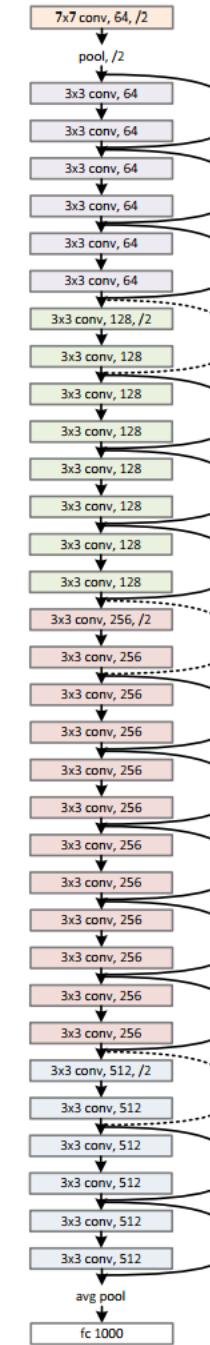
Network “Design”

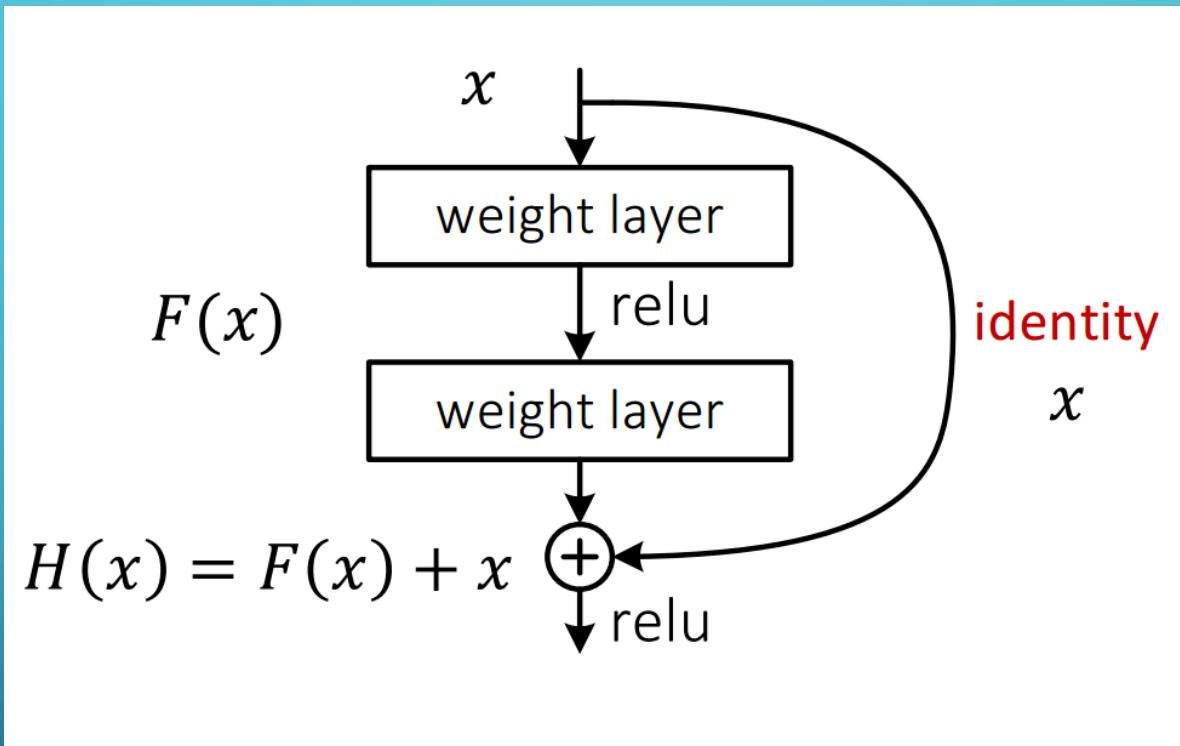
- Keep it simple
- Our basic design (VGG-style)
 - all 3x3 conv (almost)
 - spatial size /2 => # filters x2 (~same complexity per layer)
 - **Simple design; just deep!**
- Other remarks:
 - no hidden fc
 - no dropout

plain net



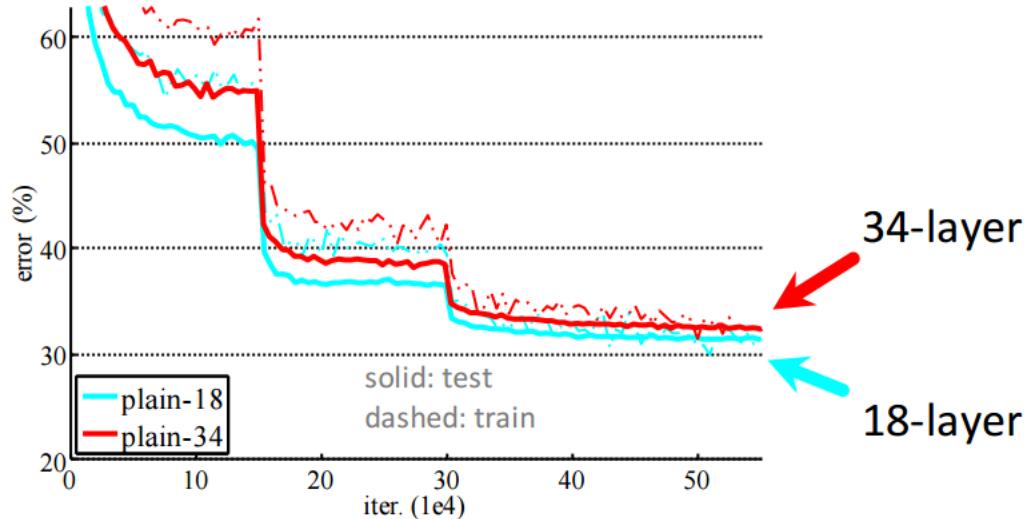
ResNet



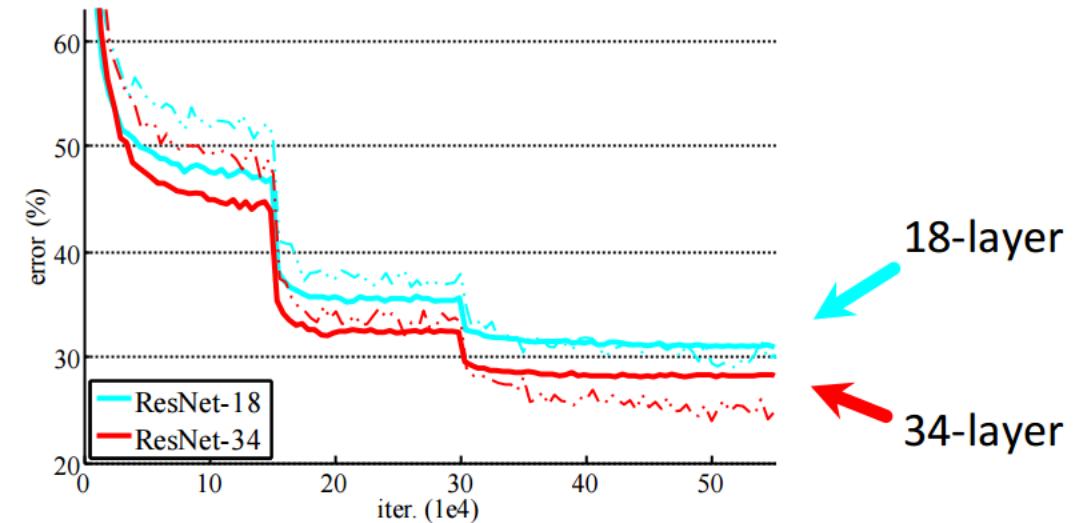


ImageNet experiments

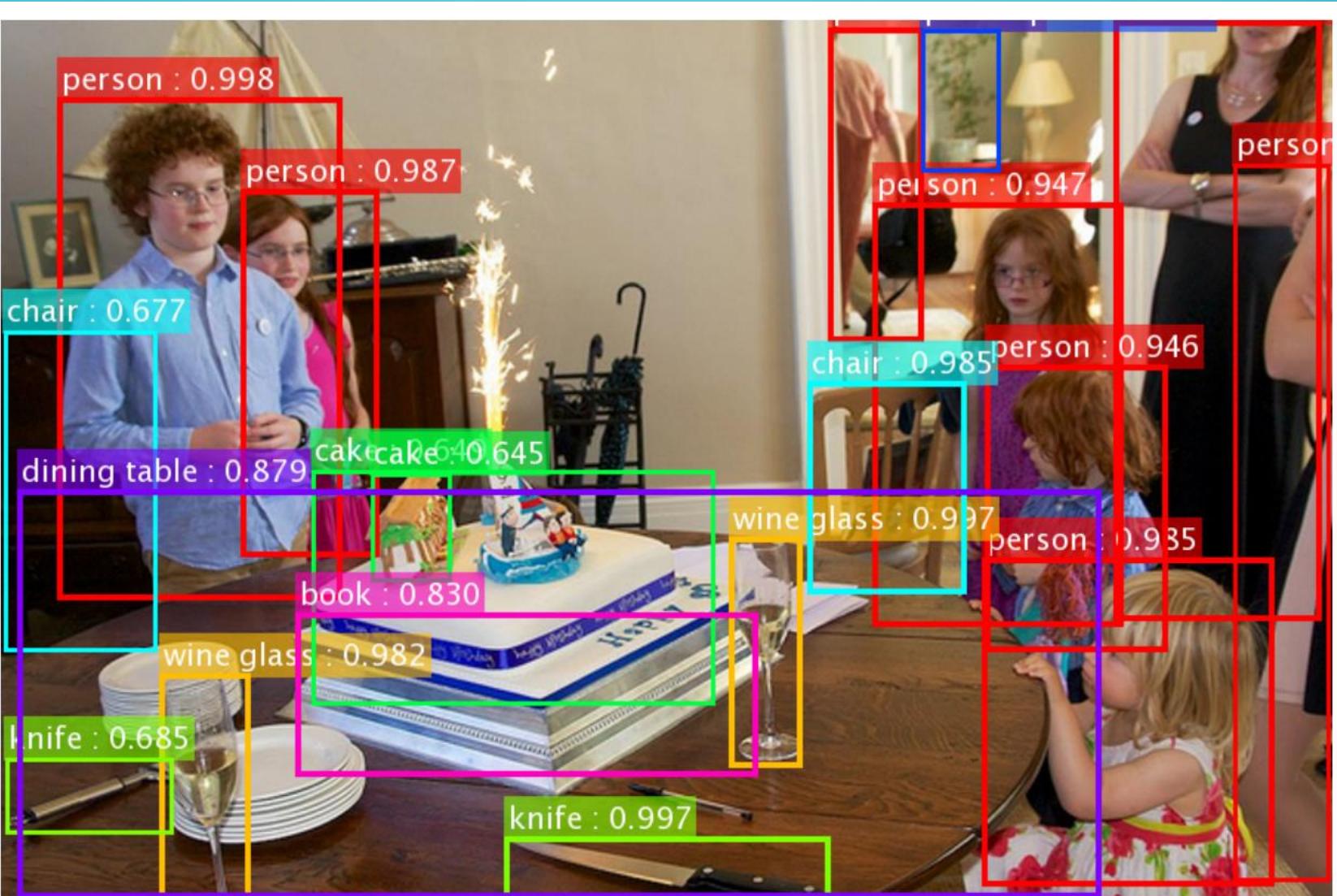
ImageNet plain nets



ImageNet ResNets



- Deep ResNets can be trained without difficulties
- Deeper ResNets have **lower training error**, and also lower test error



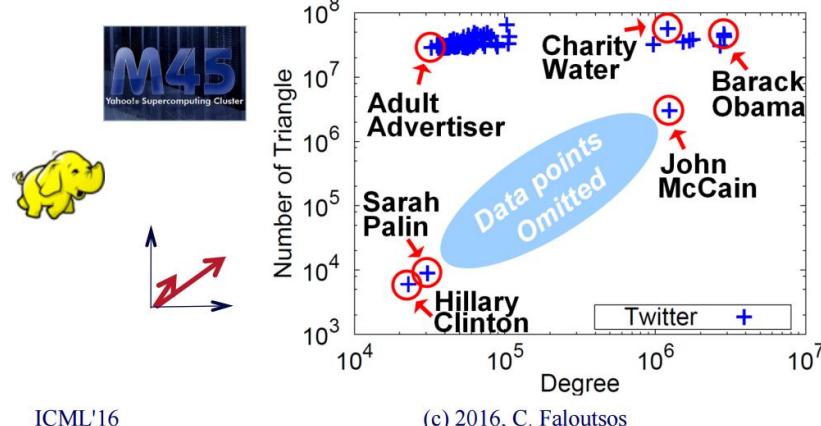
ResNet's object detection result on COCO

INVITED SPEAKERS

- “Mining Large Graphs: Patterns, Anomalies and Fraud Detection” – Ch. Faloutsos
 - http://icml.cc/2016/tutorials/faloutsos_ICML_2016.pdf

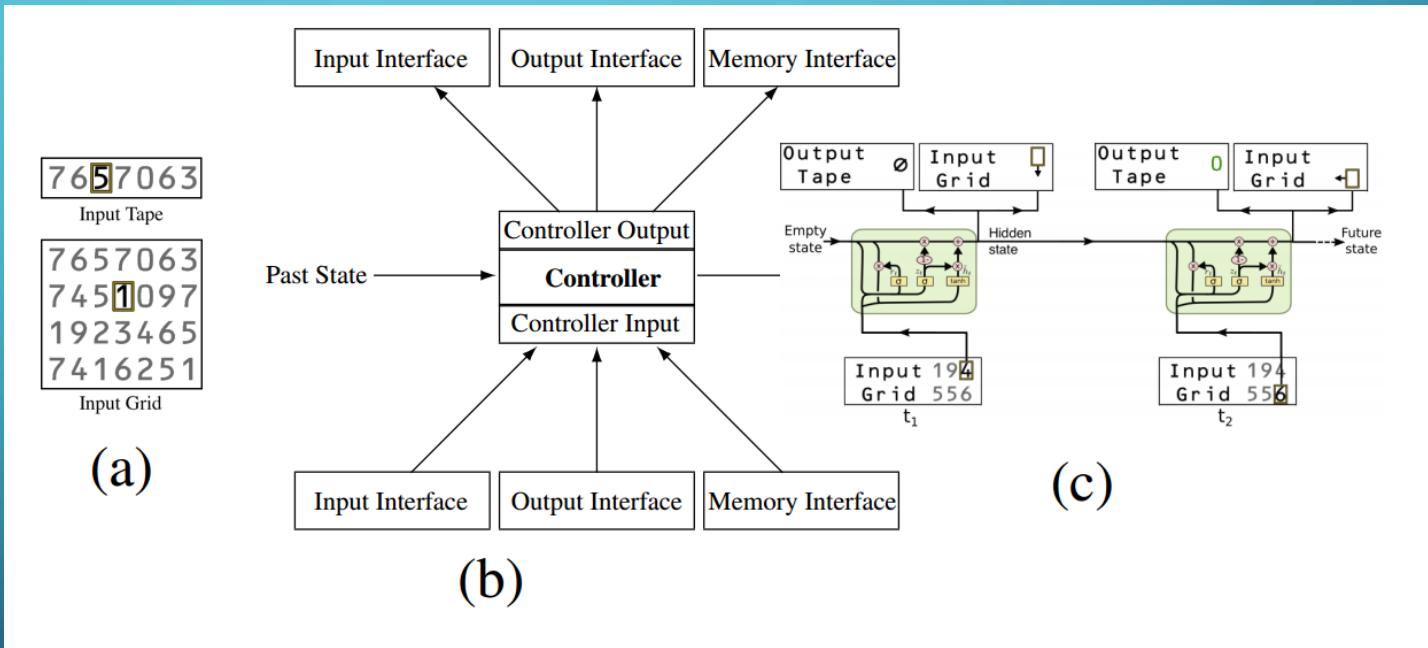
CONCLUSION#1 – Big data

- Patterns  Anomalies
- Large datasets reveal patterns/outliers that are invisible otherwise



CONFERENCE – HIGHLIGHTS

- “Learning Simple Algorithms from Examples” – W.Zaremba, T.Mikolov, A.Joulin, R.Fergus, OpenAI



- <https://www.youtube.com/watch?v=GVe6kfJnRAw>

CONFERENCE – HIGHLIGHTS

- A lot of mathematics! (Convex optimization, Learning Theory, LargeScale Classification, Kernel Methods)
- A lot of engineering ☺ (Deep neural networks, Reinforcement Learning)

WORKSHOP – HIGHLIGHTS



Neural Nets **Back to the Future** @ ICML 16

June 23rd 2016 at Crowne Plaza in NYC

A workshop **linking** the past, present and future research on neural networks

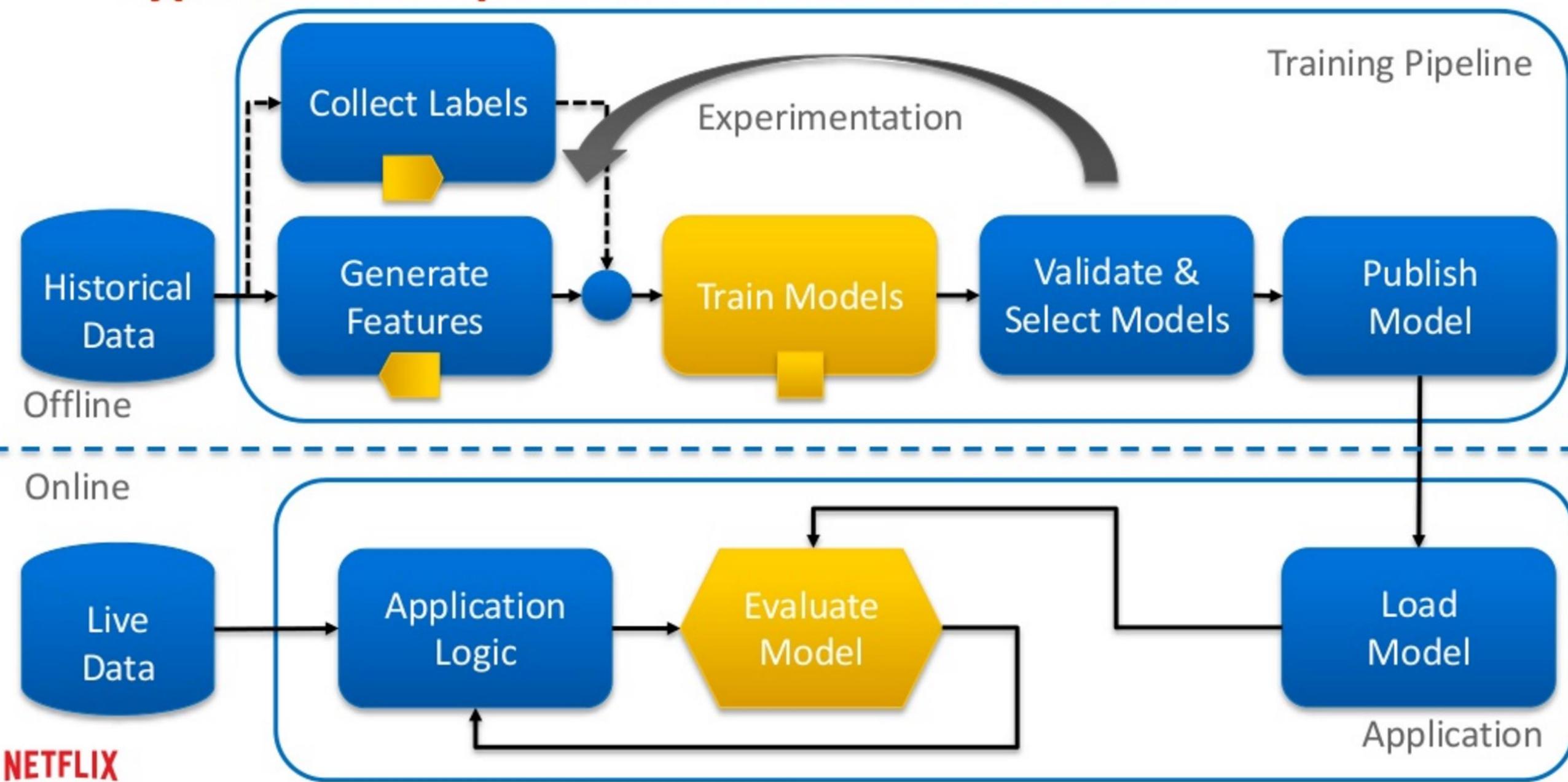
- “Let your network grow” – John Platt
- “Recurrent Neural Networks” – C. Lee Giles
- “Learning Long Term Dependencies with Gradient Descent is Difficult” – Y. Bengio

WORKSHOPS – HIGHLIGHTS

- “Real-world Machine Learning design patterns” – J.Basilico, Netflix



“Typical” ML Pipeline: A tale of two worlds



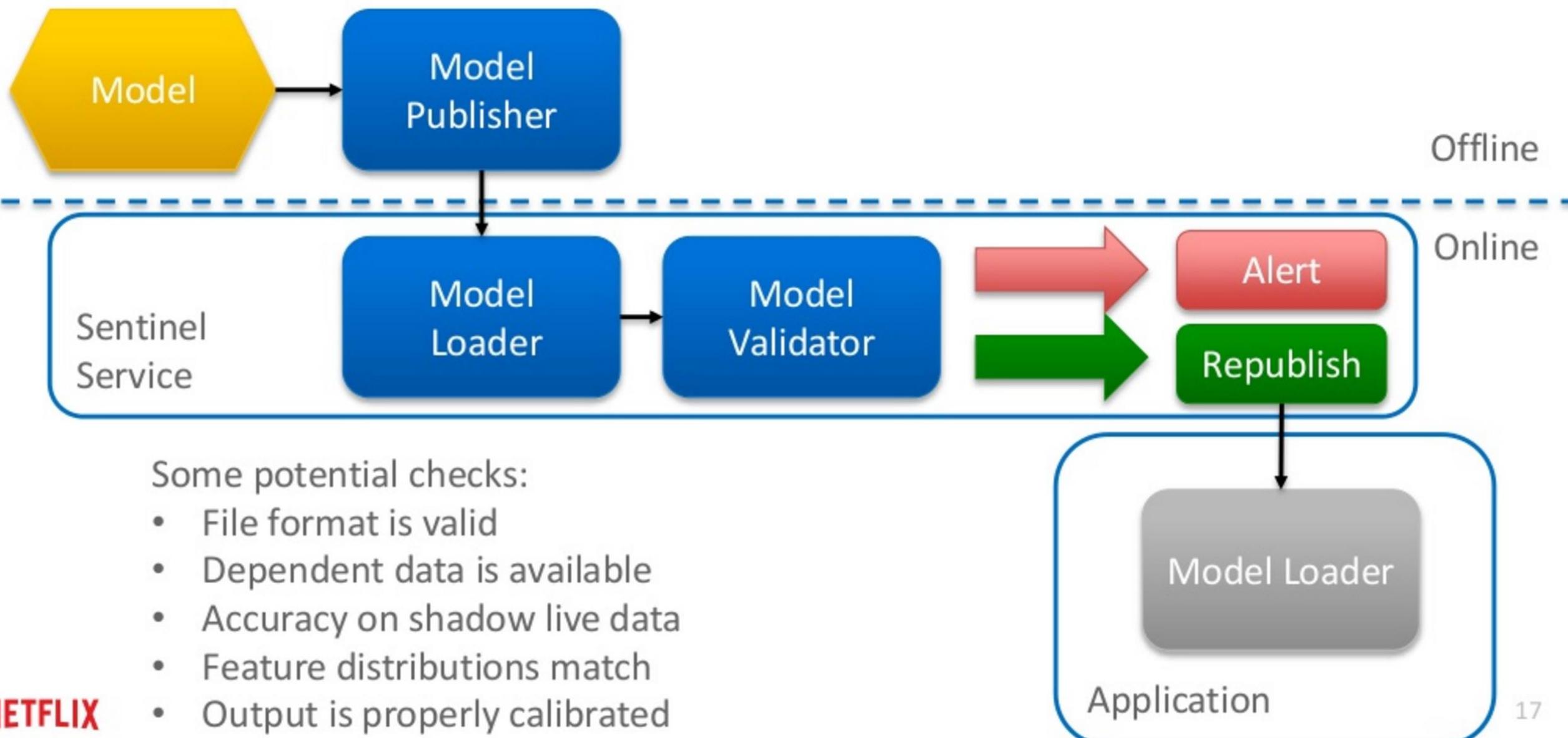
The Sentinel

“You shall not pass!”

Validate model/data in
online environment **before**
letting it go live



Sentinel: Structure

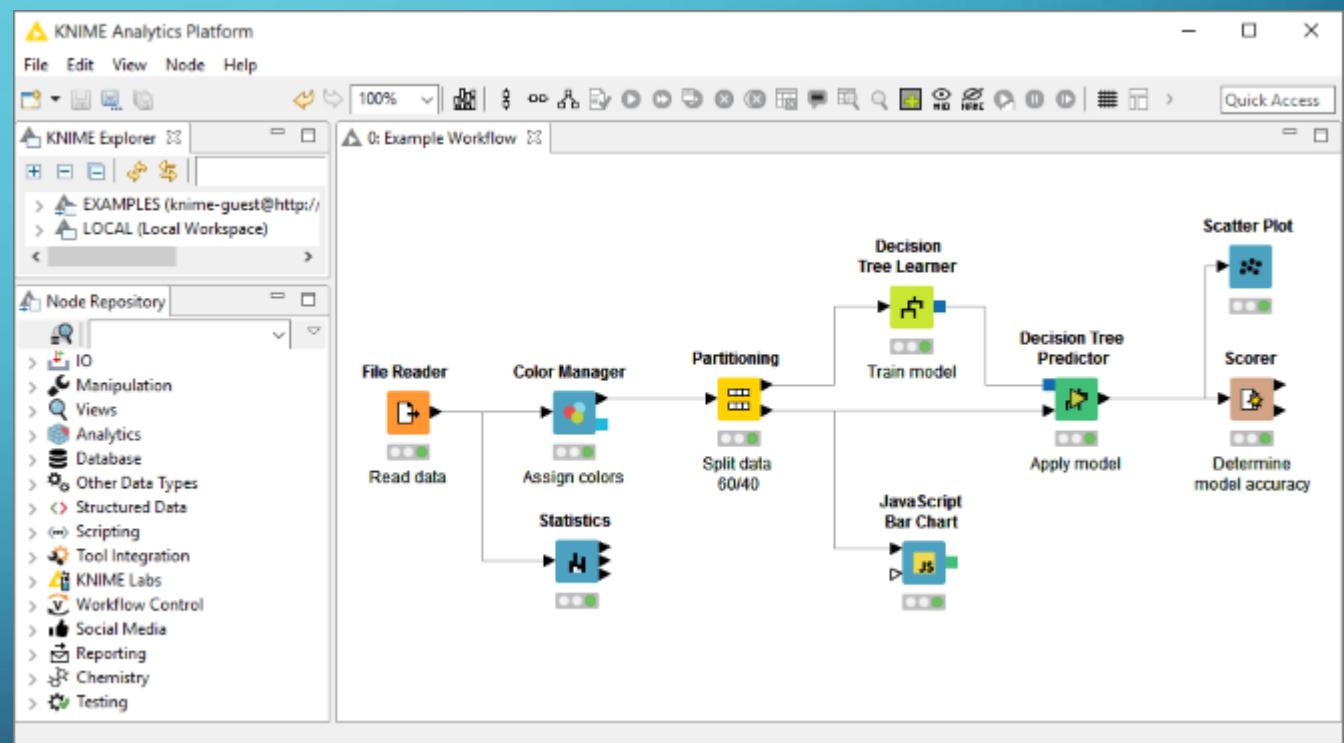


Sentinel

- Example: Checking that new ranking model is valid and performs better than previous one
- Pros:
 - Using a model requires both code and data are available
 - Models may need to be versioned along-side code changes
 - Ensure that a new model is no worse than previous one
- Cons:
 - Sentinel needs to be in sync with application code
 - Difficult to choose failure thresholds for data-based checks

WORKSHOPS – HIGHLIGHTS

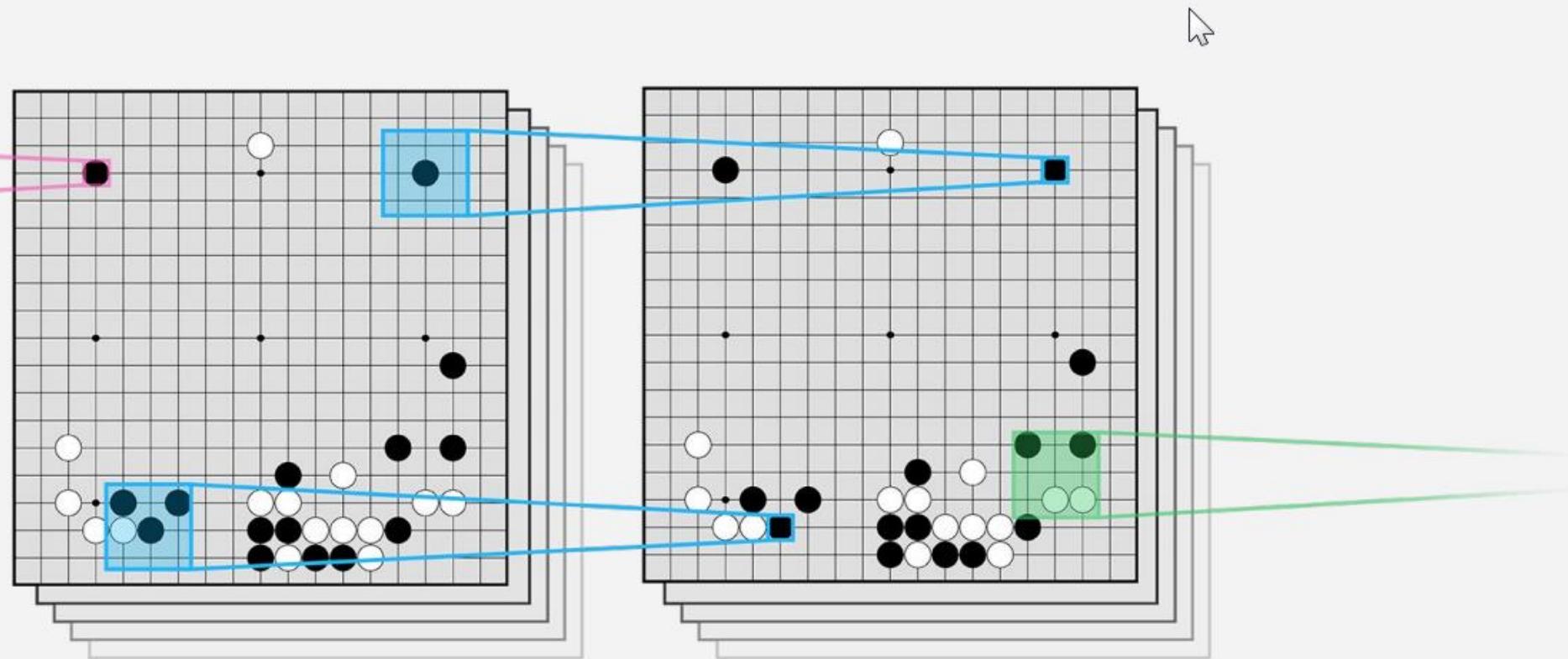
- “Modular Machine Learning with KNIME”
- “Probabilistic Demand Forecasting at Amazon Scale”
- “Intel® MKL-DNN – an Open-Source Library of Neural Network Primitives”



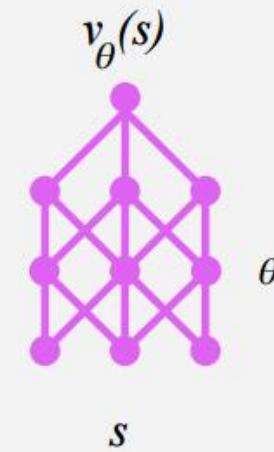
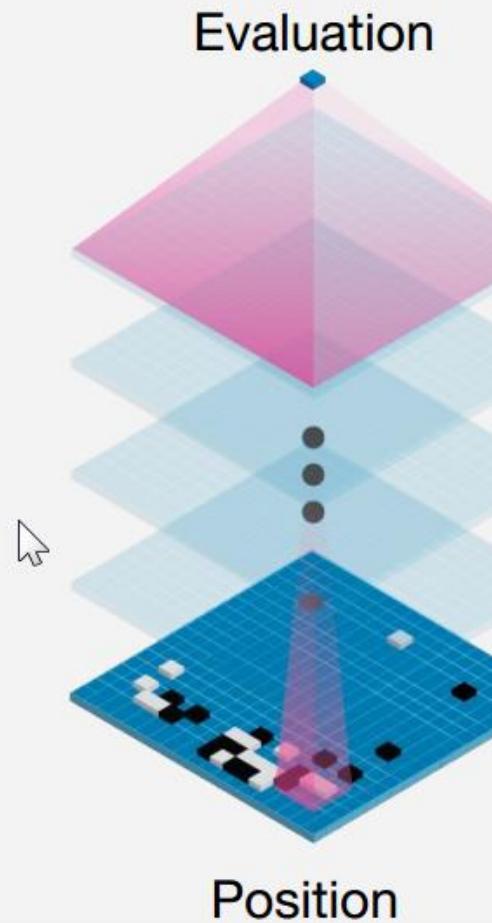
ALPHA GO



Convolutional neural network

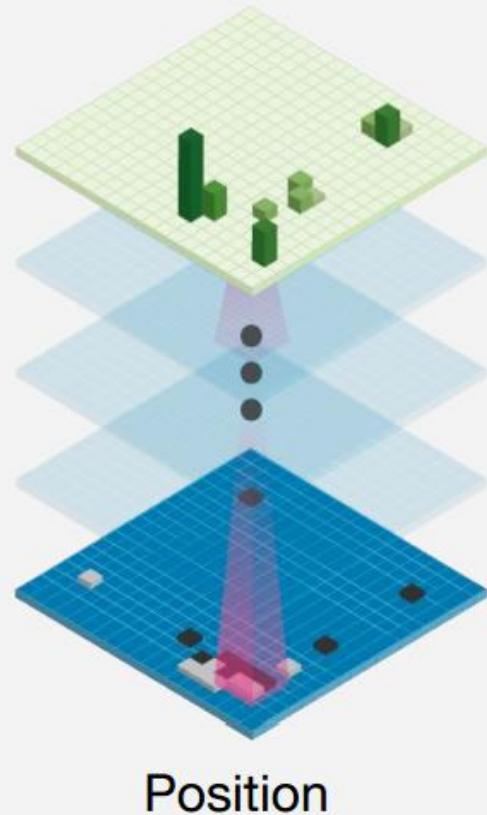


Value network



Policy network

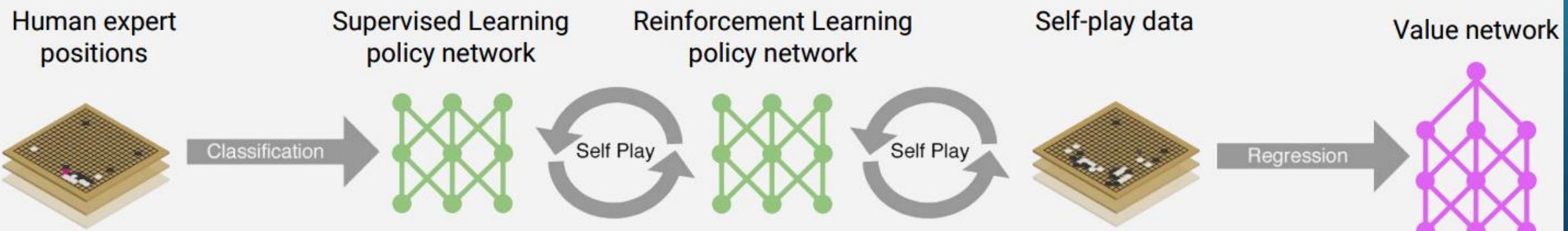
Move probabilities



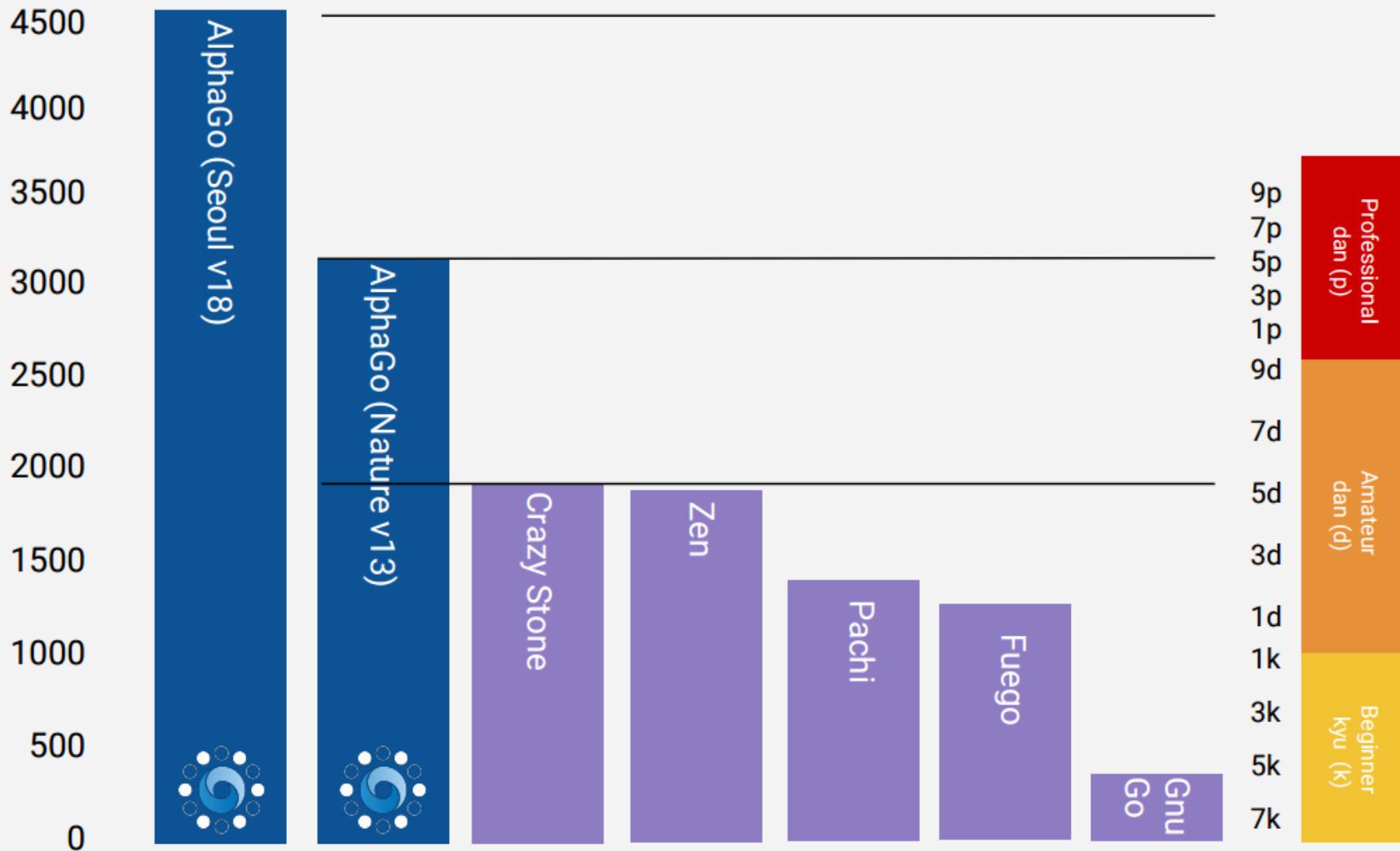
$$p_{\sigma}(a|s)$$


A diagram of a neural network structure. On the left, a blue square labeled s represents the input state. On the right, a green square labeled σ represents the output action probability distribution. The network consists of two layers of green nodes connected by green lines. The input s is processed by the first layer, which then produces the output $p_{\sigma}(a|s)$.

Neural network training pipeline



Evaluating AlphaGo against computers



LINKS

- <https://icml.cc>
- <http://icml.cc/2016/wp-content/uploads/ICML-Book-Web-Version.pdf>
- <http://jmlr.org/proceedings/papers/v48/>
- <https://openai.com/blog/>
- <https://gogameguru.com/i/2016/03/deepmind-mastering-go.pdf>
- <http://www.nature.com/nature/journal/v529/n7587/full/nature16961.html>
- <http://jmlr.org/proceedings/papers/v48/zaremba16.pdf>
- <https://arxiv.org/pdf/1512.02595v1.pdf>
- <http://www.slideshare.net/justinbasilico/is-that-a-time-machine-some-design-patterns-for-real-world-machine-learning-systems>



The End

