

# **Redes Neuronales**

Temas avazandos

# Agenda

- Autoencoder
- Generative Adversarials Networks
- Transfer Learning
- Quantization
- Sparsification
- Conclusiones



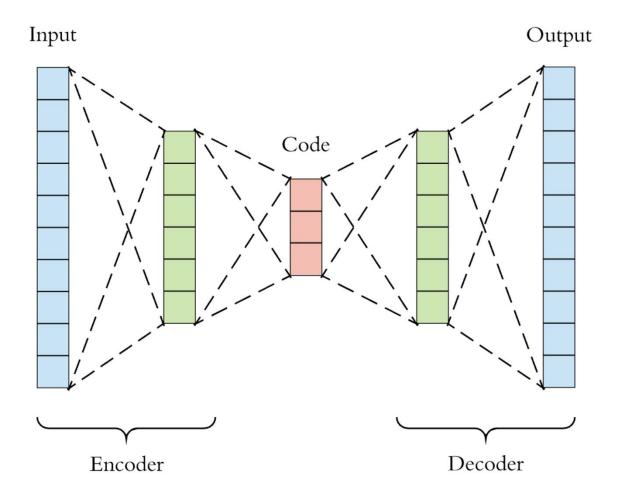
#### **Autoencoders**

#### Autoencoders son útiles para:

- Reducción de dimensionalidad
- Reducción de ruido
- Generación automática
- Recomendación

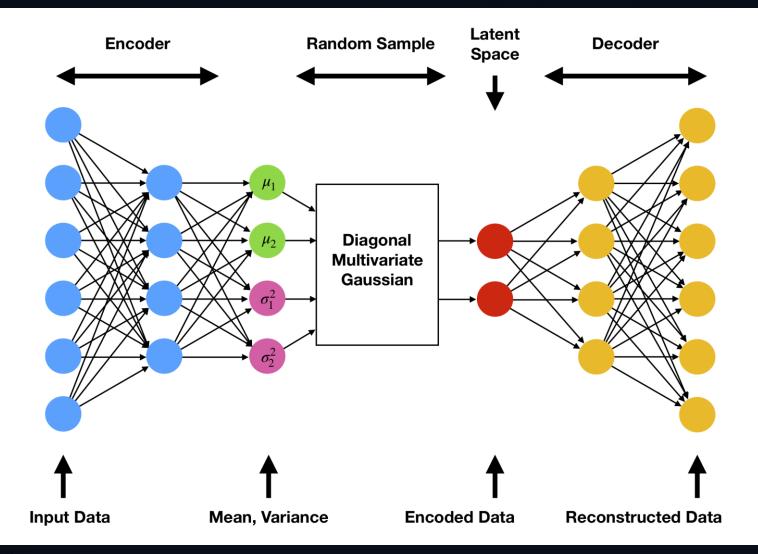


### **Autoencoders**





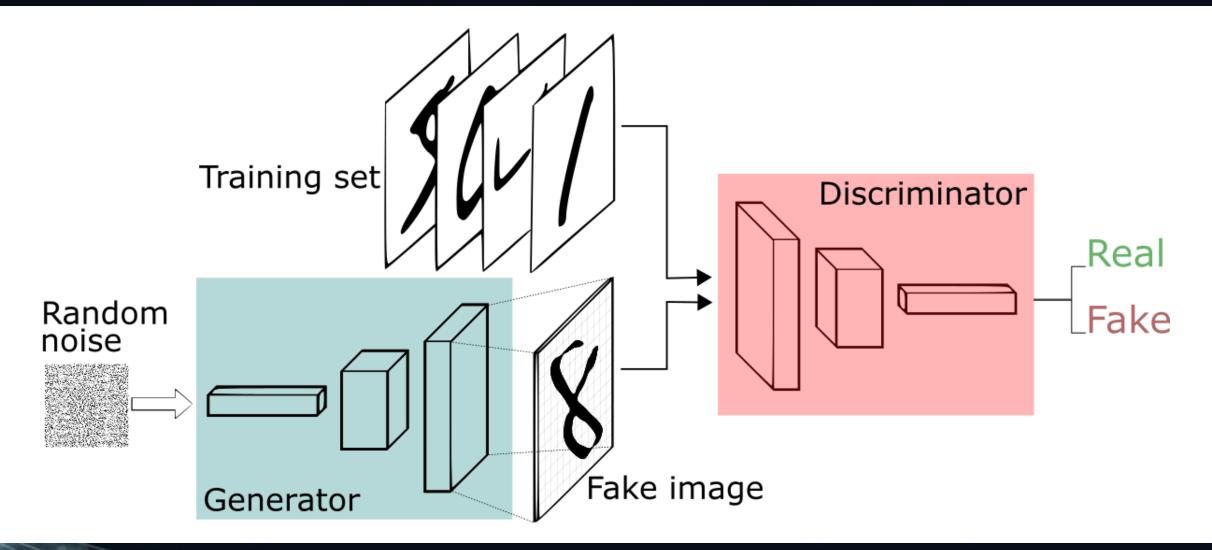
#### Autoencoders



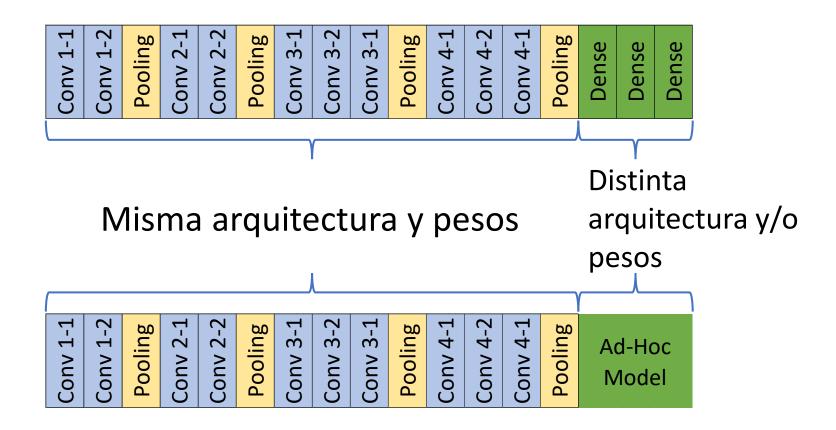




#### **GANs**

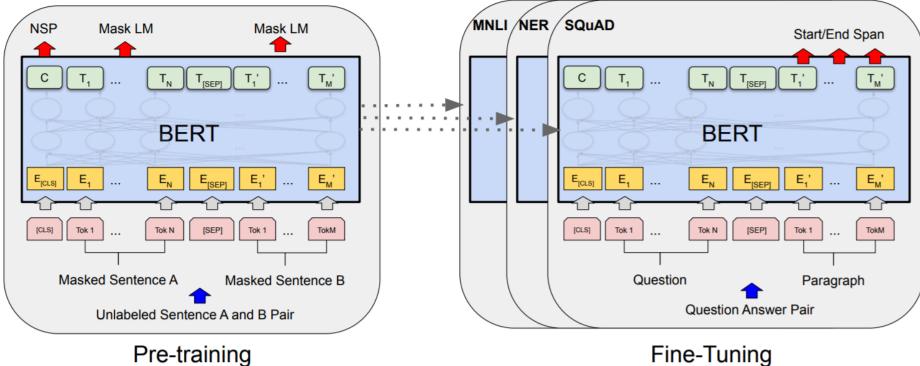


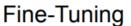
#### Transfer learning





### **Transfer learning: BERT**



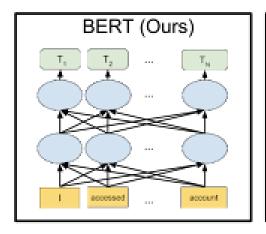


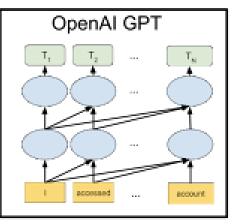
BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding. Jacob Devlin, Ming-Wei Chang, Kenton Lee, Kristina Toutanova https://arxiv.org/abs/1810.04805

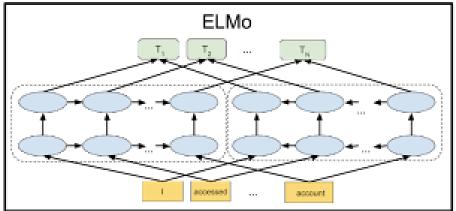




# Transfer learning: ¡más modelos!





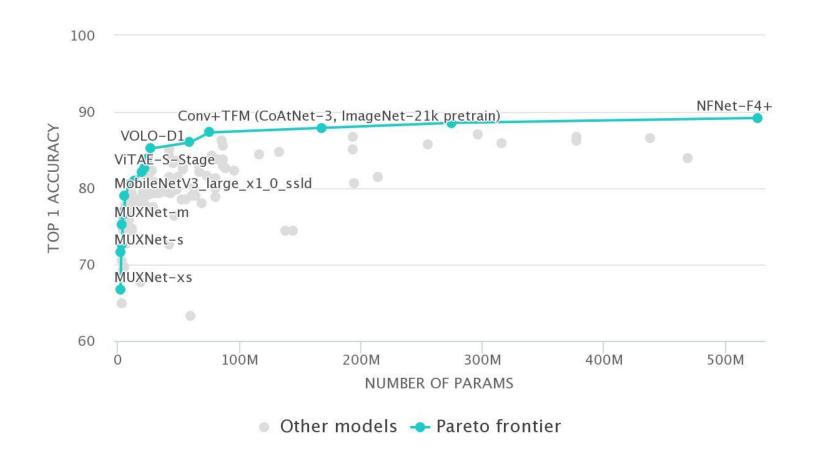


https://ai.googleblog.com/2018/11/open-sourcing-bert-state-of-art-pre.html





### Transfer learning: + datos + parámetros!

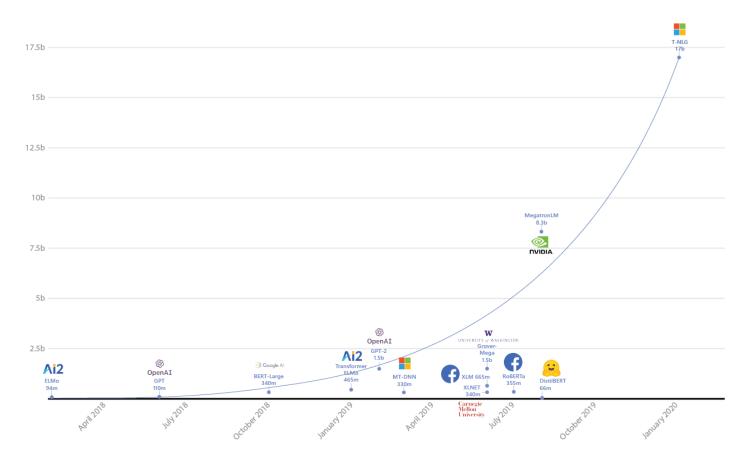


 $https://papers with code.com/sota/image-classification-on-imagenet? dimension=Number \% 20 of \% 20 params \& tag\_filter=0. \\$ 





# Transfer learning: + datos + parámetros!

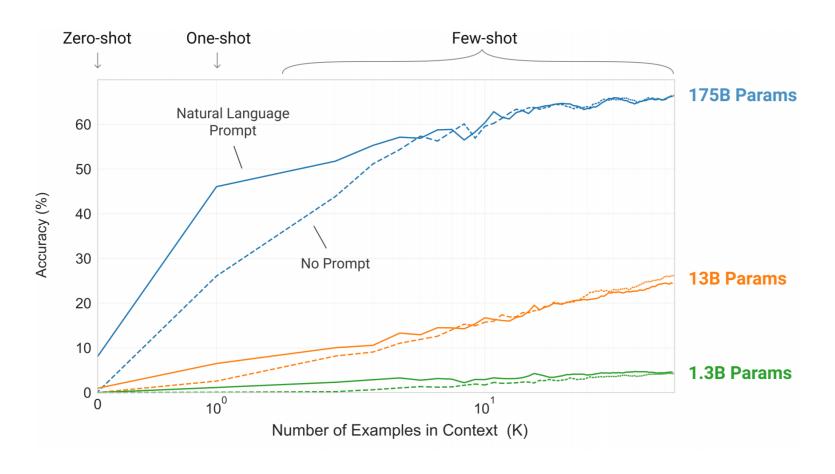


https://www.microsoft.com/en-us/research/blog/turing-nlg-a-17-billion-parameter-language-model-by-microsoft/





# Transfer learning: + datos + parámetros!



Brown, T. B., Mann, B., Ryder, N., Subbiah, M., Kaplan, J., Dhariwal, P., ... & Agarwal, S. (2020). Language models are few-shot learners. arXiv preprint arXiv:2005.14165...





# Quantization

Technique	Data requirements	Size reduction	Accuracy	Supported hardware
Post-training float16 quantization	No data	Up to 50%	Insignificant accuracy loss	CPU, GPU
Post-training dynamic range quantization	No data	Up to 75%	Accuracy loss	CPU, GPU (Android)
Post-training integer quantization	Unlabelled representative sample	Up to 75%	Smaller accuracy loss	CPU, GPU (Android), EdgeTPU, Hexagon DSP
Quantization-aware training	Labelled training data	Up to 75%	Smallest accuracy loss	CPU, GPU (Android), EdgeTPU, Hexagon DSP

https://www.tensorflow.org/lite/performance/model\_optimization#quantization





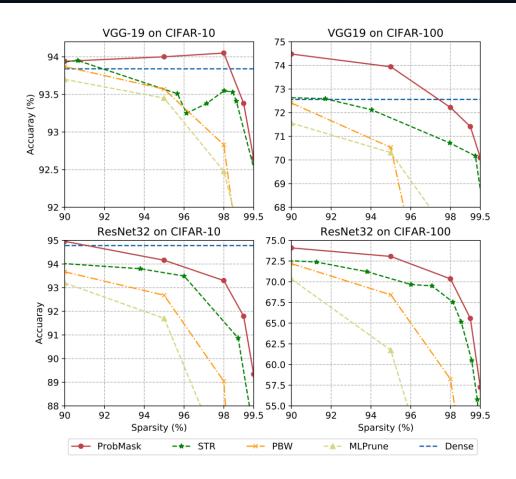
# Quantization

Model	Top-1 Accuracy (Original)	Top-1 Accuracy (Post Training Quantized)	Top-1 Accuracy (Quantization Aware Training)	Latency (Original) (ms)	Latency (Post Training Quantized) (ms)	Latency (Quantization Aware Training) (ms)	Size (Original) (MB)	Size (Optimized) (MB)
Mobilenet-v1- 1-224	0.709	0.657	0.70	124	112	64	16.9	4.3
Mobilenet-v2- 1-224	0.719	0.637	0.709	89	98	54	14	3.6
Inception_v3	0.78	0.772	0.775	1130	845	543	95.7	23.9
Resnet_v2_101	0.770	0.768	N/A	3973	2868	N/A	178.3	44.9

 $https://www.tensorflow.org/lite/performance/model\_optimization \# quantization$ 



# **Spasification**

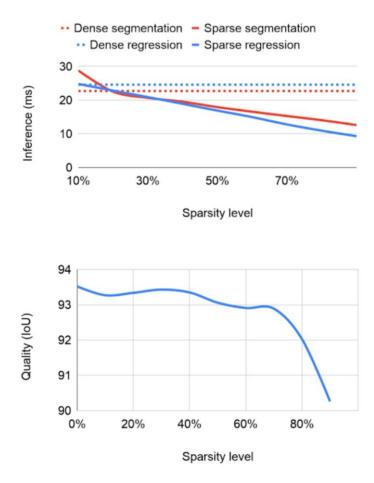


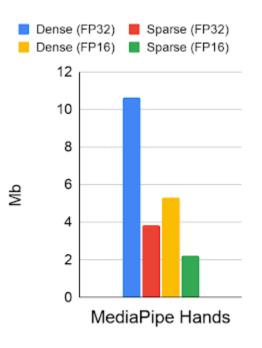
https://openaccess.thecvf.com/content/CVPR2021/papers/Zhou\_Effective\_Sparsification\_of\_Neural\_Networks\_With\_Global\_Sparsity\_Constraint\_CVPR\_2021\_paper.pdf

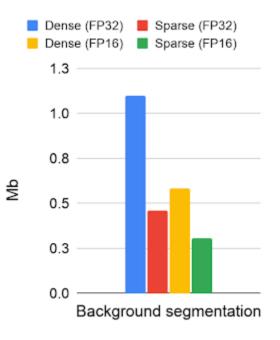




# **Spasification**







https://ai.googleblog.com/2021/03/accelerating-neural-networks-on-mobile.html





# Herramientas y más...

- Tensorflow
- Pytorch
- Apache mxnet
- ONNX
- HuggingFace
- NVIDIA TensorRT
- NVIDIA Triton Inference Server
- Tensorflow Serving
- Tensorflow Lite
- ...







