



# From Scratch to Large Pre-Trained Models. A Comparative Study for Medical Image Classification

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#### **3rd Italian Conference on Big Data and Data Science**





## Introduction

- The use of AI in healthcare fields is constantly increasing.
- Diagnostic technologies such as x-ray, Magnetic Resonance Imaging (MRI) and computerized tomography (CT) use images to represent information.
- 90% of data in healthcare are images.
- Image classification is a key task in healthcare, with applications in a variety of areas, including:
  - Diagnosis
  - Treatment planning
  - Research

Study of
Convolutional
Neural Network
(CNN) and Vision
Transformer (ViT)
architectures

From scratch implementation

Workflow

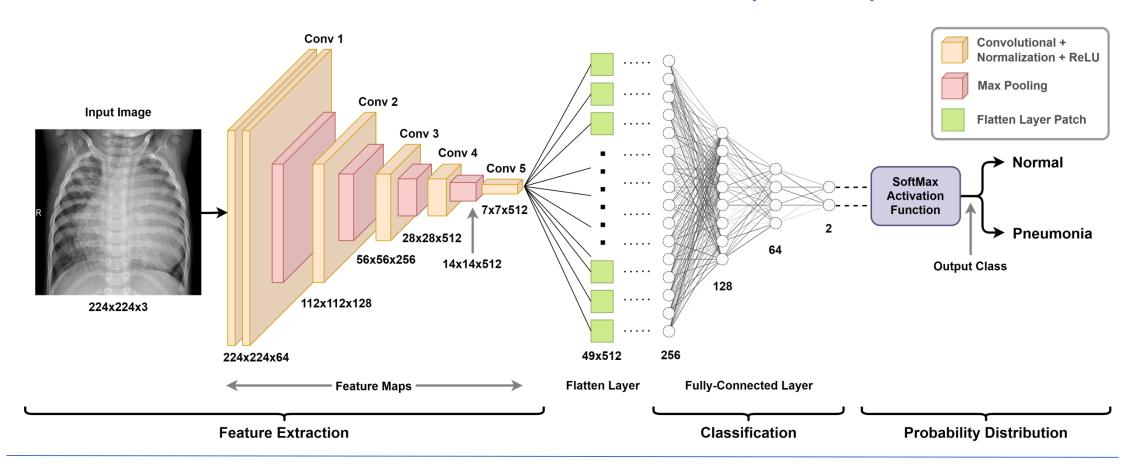
Comparison with Pre-Trained approach

Performance
evaluation
comparison
between consumer
laptop and a Cloud
HPC Infrastructure
as a Service (IaaS)





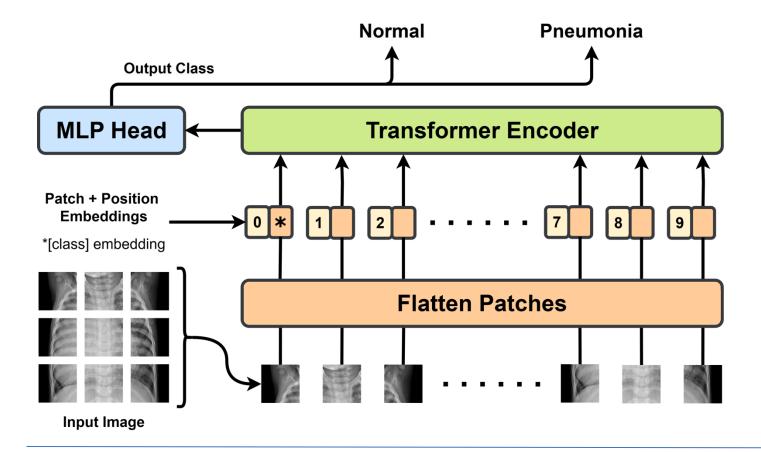
# Convolutional Neural Network (CNN)

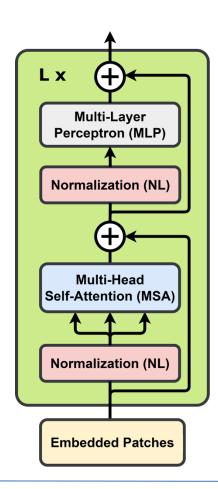






## Vision Transformer (ViT)









### CNNs vs ViTs

Feature	CNNs	ViTs
Learning patterns	Local patterns	Global patterns
Image processing	Sequential	Parallel
Accuracy	Typically lower	Can be higher
Inference times	Typically faster	Can be slower
Memory usage	Lower	Higher
Training complexity	Lower	Higher





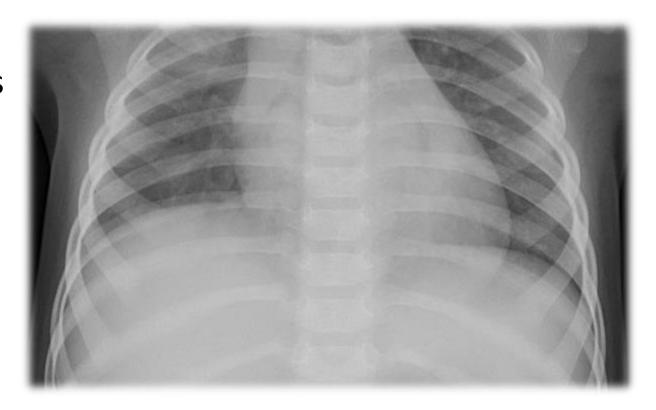
#### Dataset

The dataset selected for this study is the Chest X-Ray Images (Pneumonia)

Content: 5863 JPEG-formatted X-ray pictures

#### Classes:

- Normal
- Pneumonia

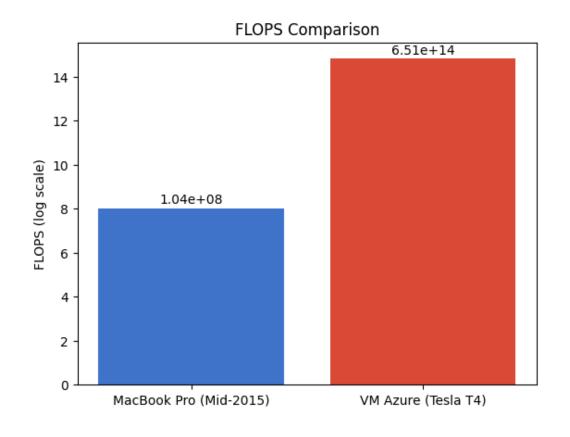






## Test beds

	Macbook Pro (Mid- 2015)	(Mid- Azure VM	
CPU	i7-4770HQ, 4 cores, 2.2Ghz	8 cores, 2.5GHz	
RAM	16Gb, DDR3	56 GB, DDR4	
GPU	Not used	Nvidia Tesla T4, 16Gb DDR6	
FLOPS	10.38M	65.13T	



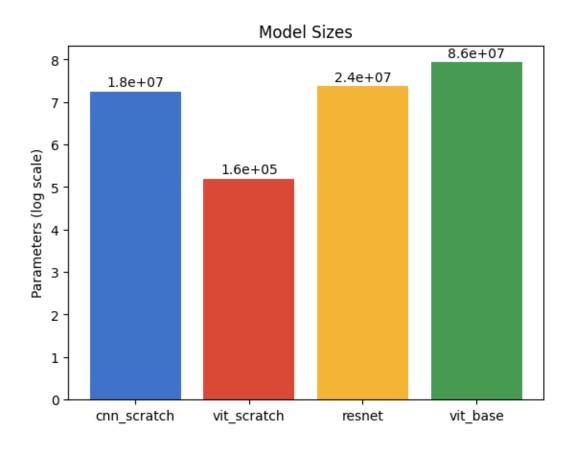




## Models Parameters

Model	Layers	Hidden size <i>D</i>	MLP size	Heads	Params
ViT_scratch	4	56	224	4	155.7K
ViT-Tiny	12	192	768	3	5.5M
ViT-Base	12	768	3072	12	86M
ViT-Large	24	1024	4096	16	307M
ViT-Huge	32	1280	5120	16	632M

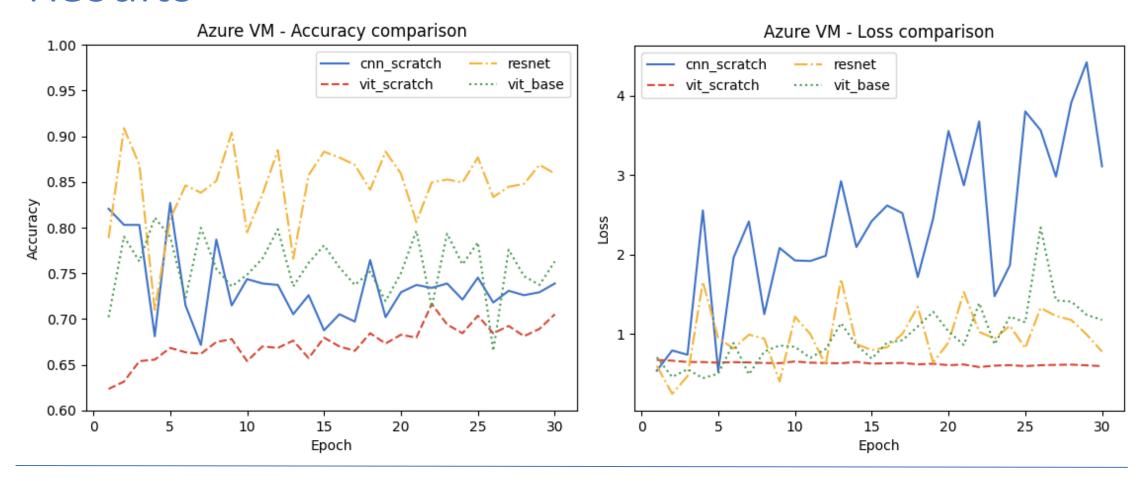
Model	Parameters	
CNN_scratch	17,6M	
ViT_scratch	155K	
ResNet50	23,5M	
ViT_Base	86M	







## Results

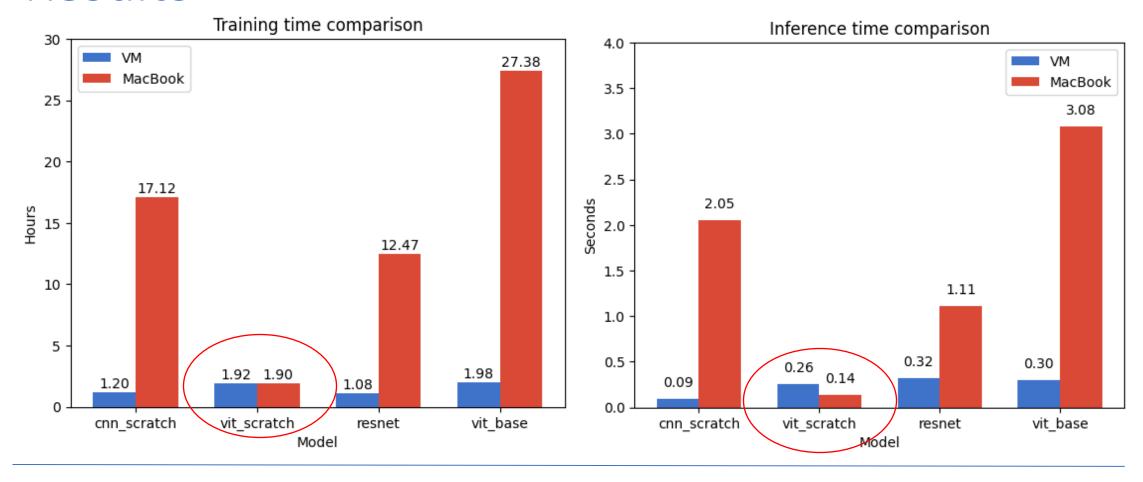


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## Results



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## Conclusion

- Transfer Learning is particularly useful in domains where data availability is scarce
- HPC environment significantly accelerates the development process of large deep learning models





#### **Future Works**

- Compare other CNN and ViT models.
- Explore alternative approaches such as Federated Learning to ensure clinical data privacy and parallel process-ing towards the perspective of classifying Big Medical Data in a collaborative federated healthcare scenario using HPC services.





# Thank you

**Any Question?** 



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