



醫學影像專題

第二組

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Outline

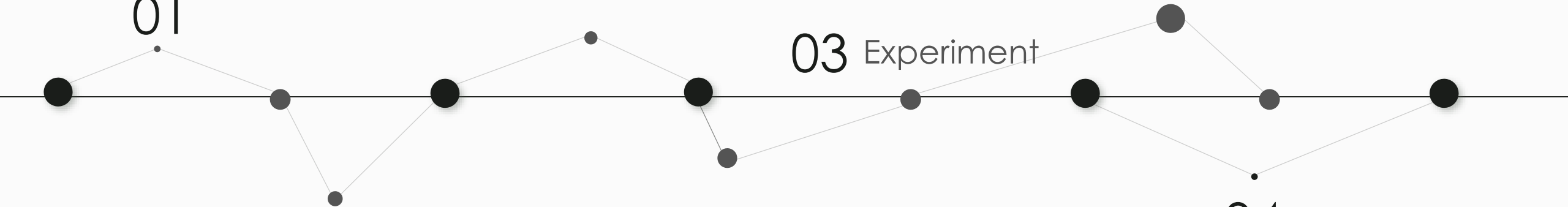
Introduction

01

03 Experiment

02 Model

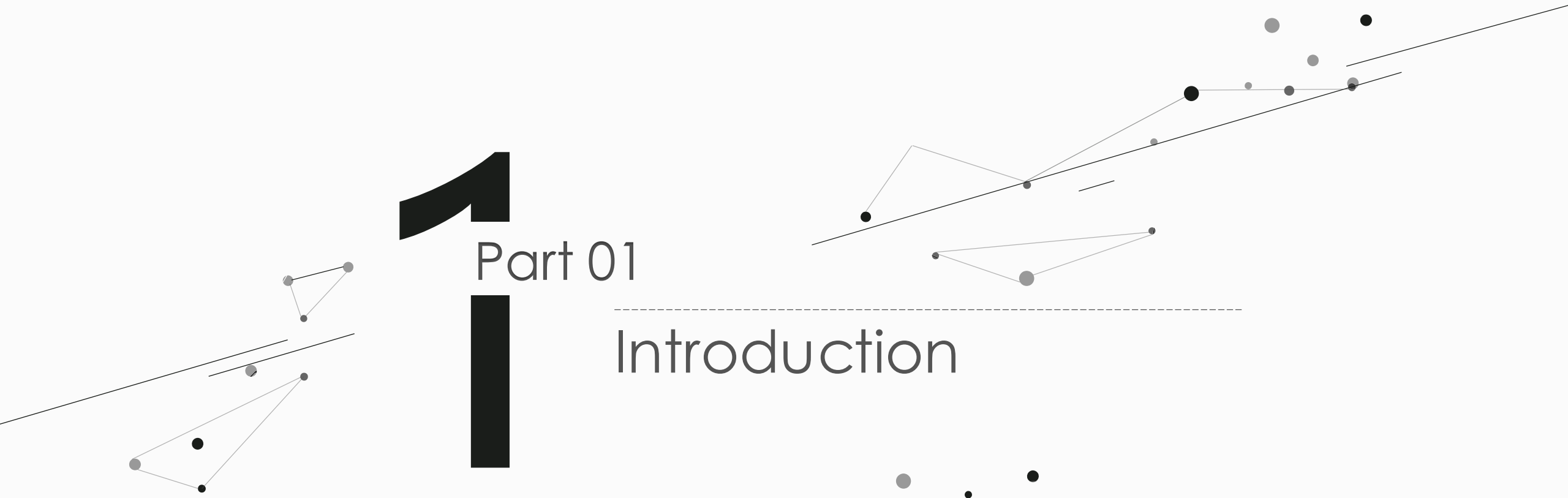
04
Conclusion

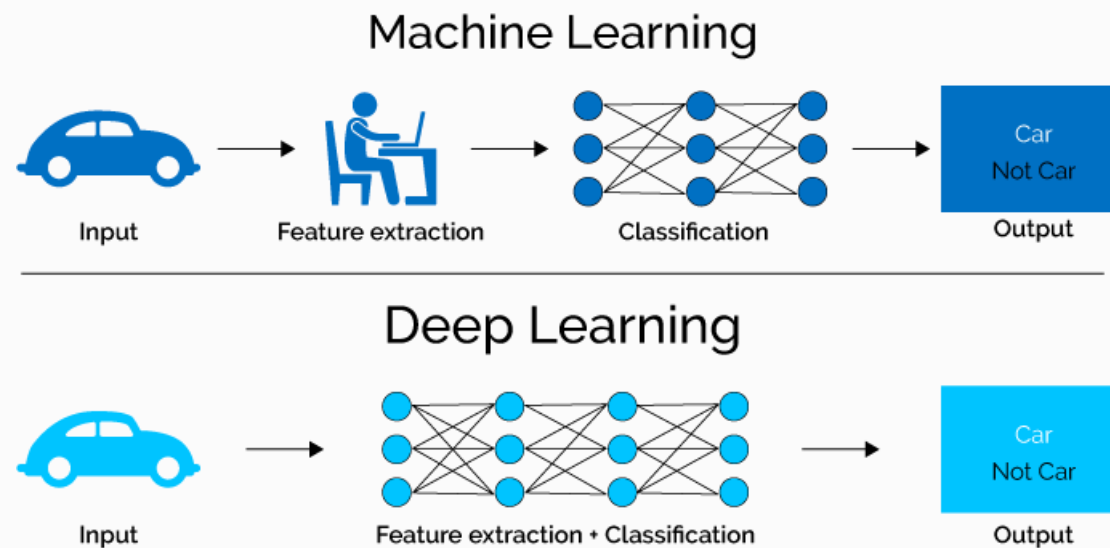
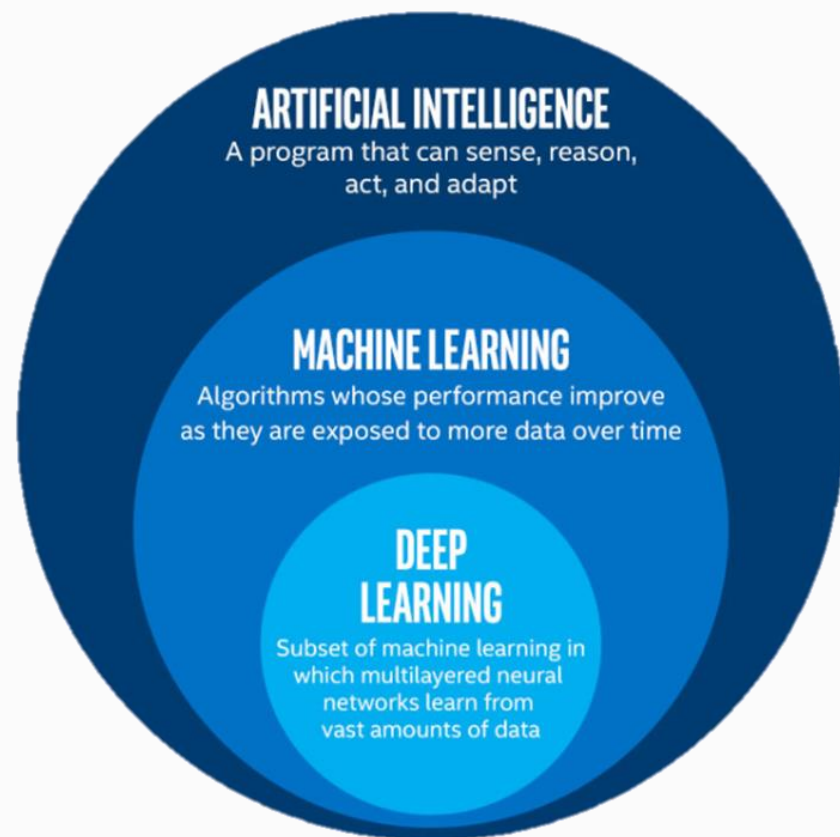


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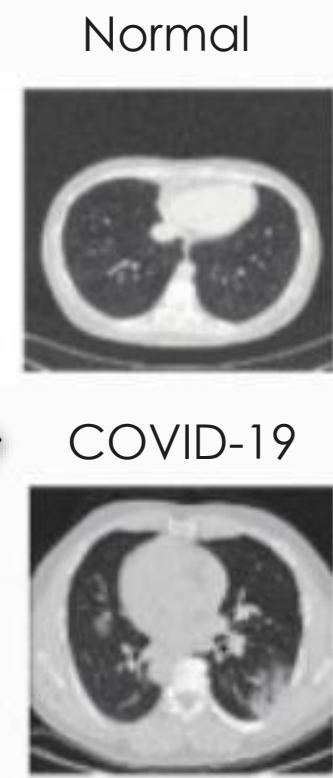
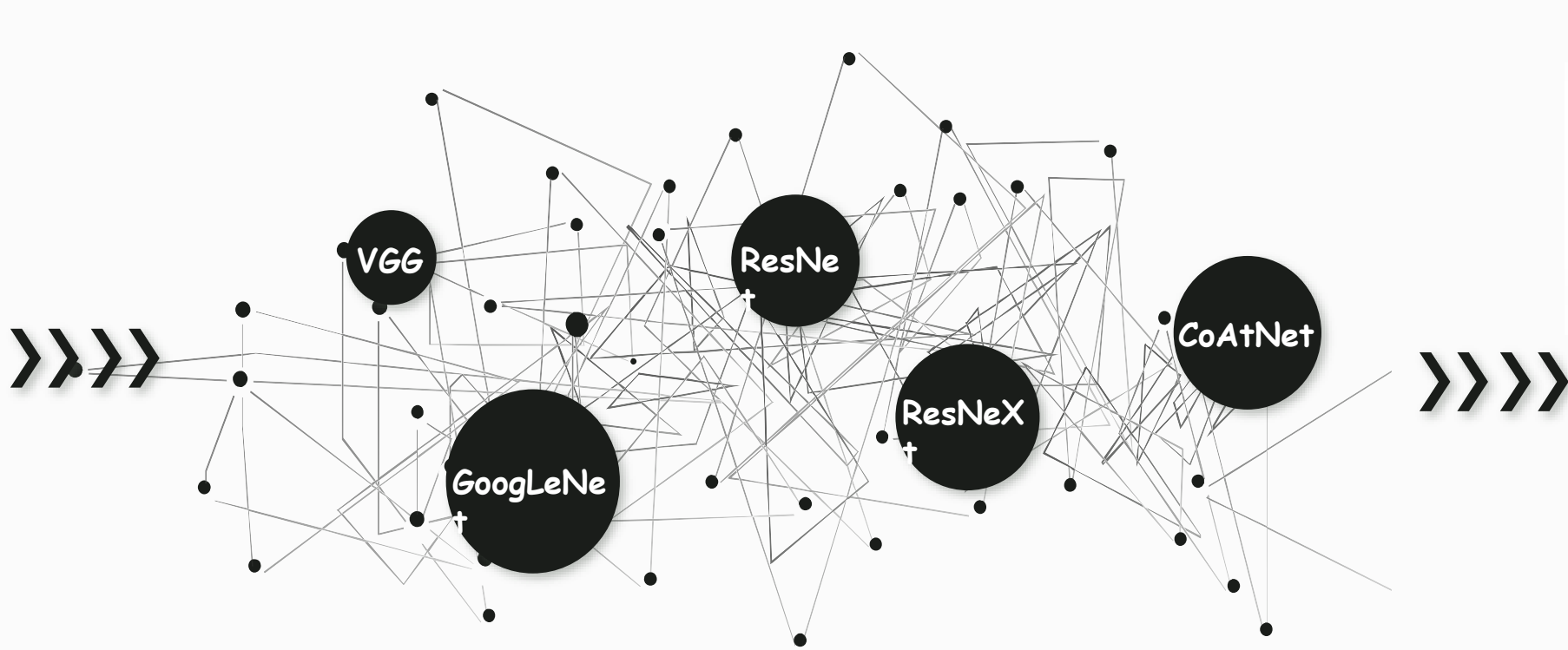
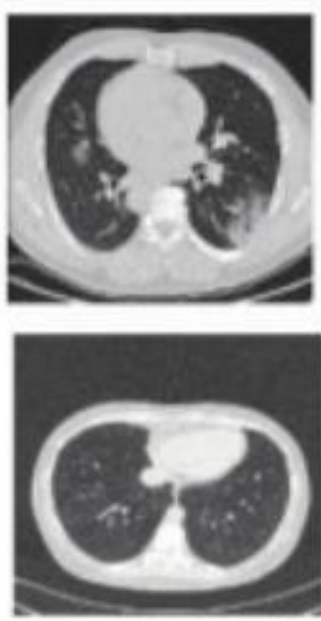
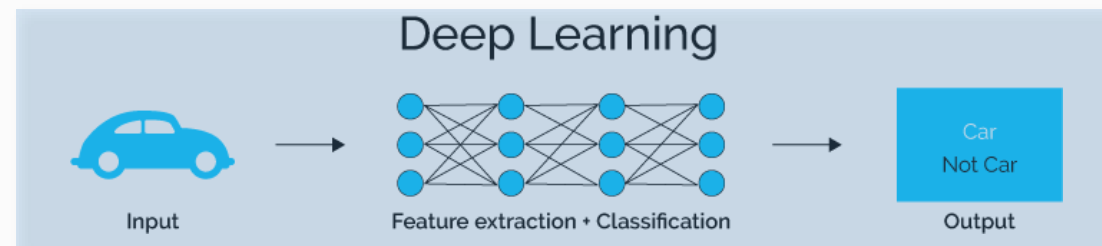
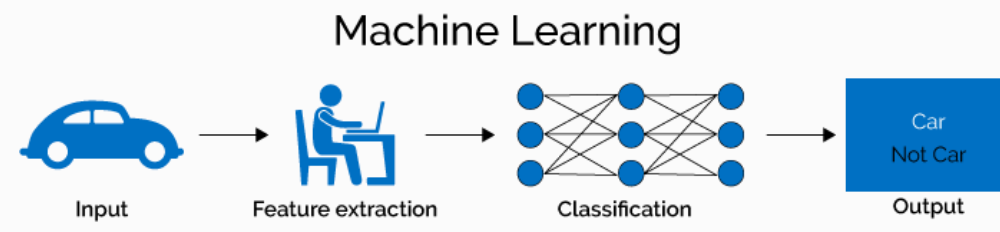
Part 01

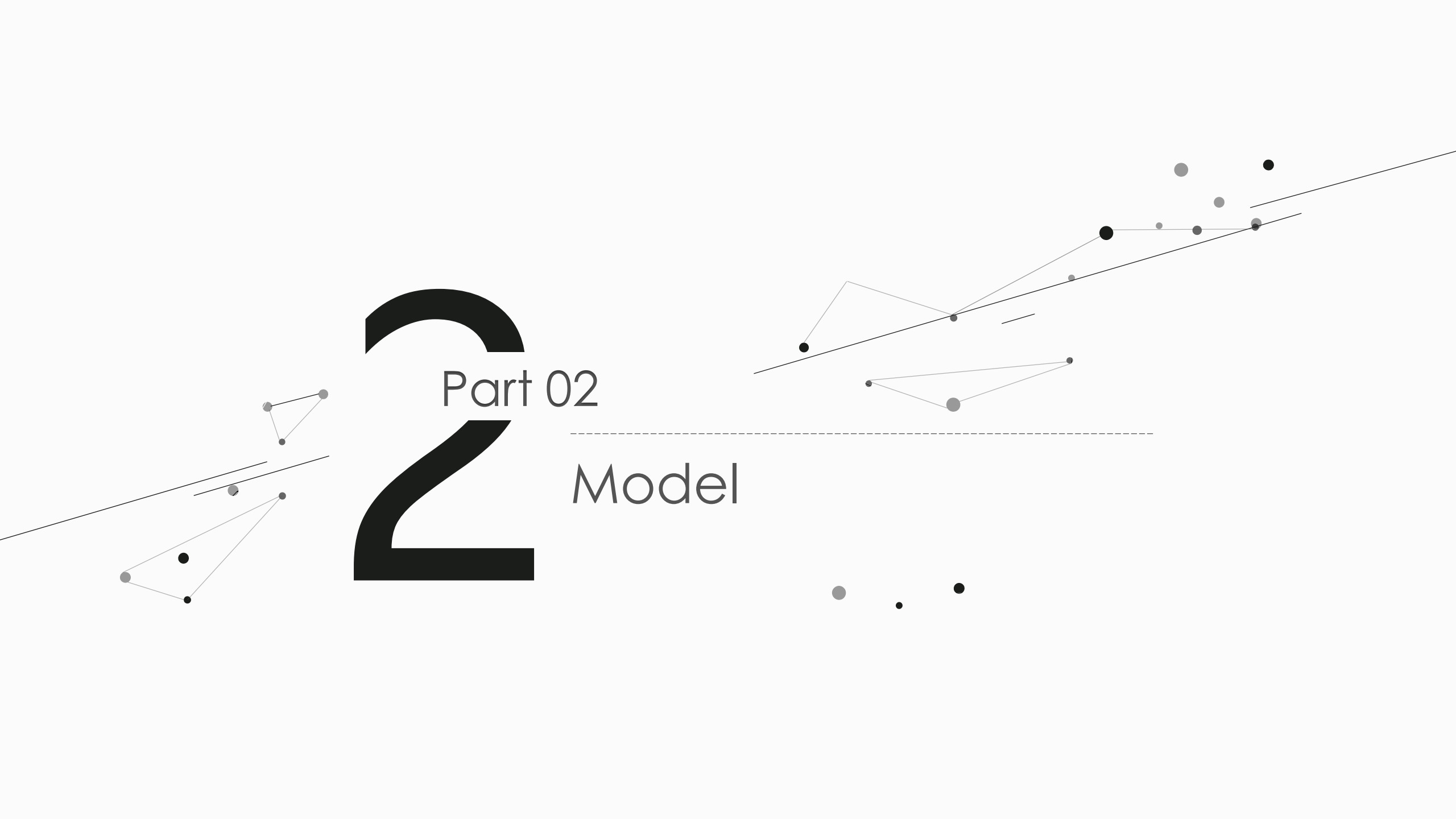
Introduction





01/



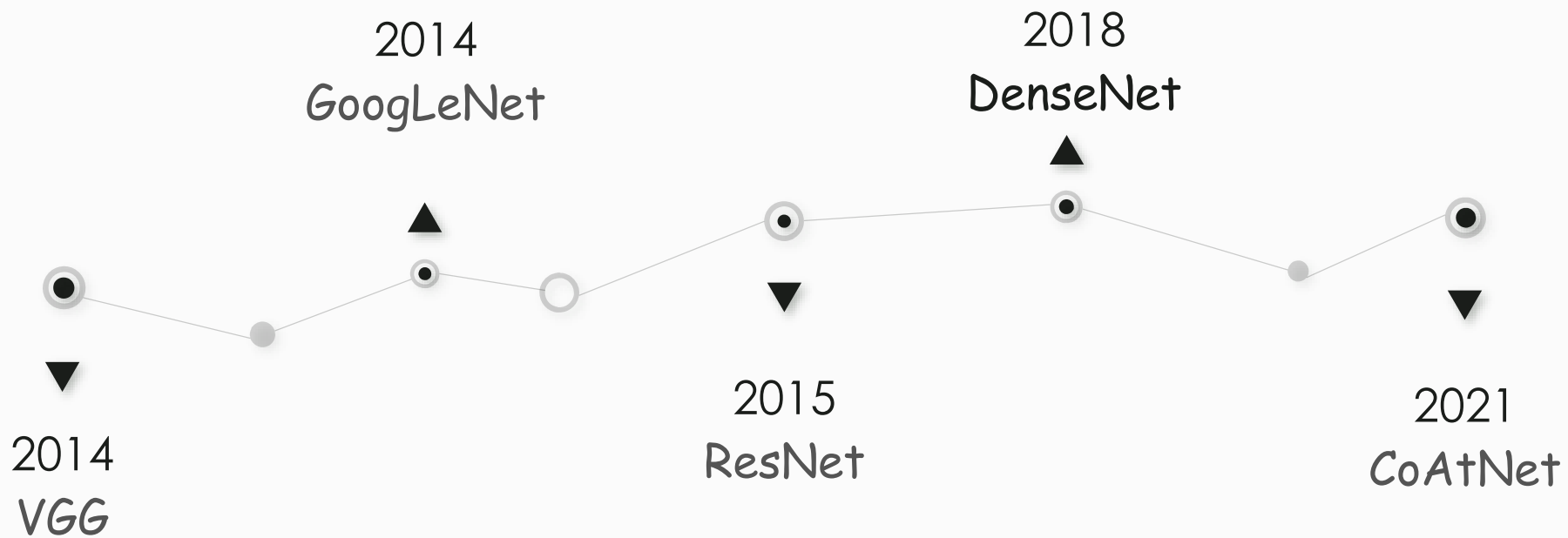


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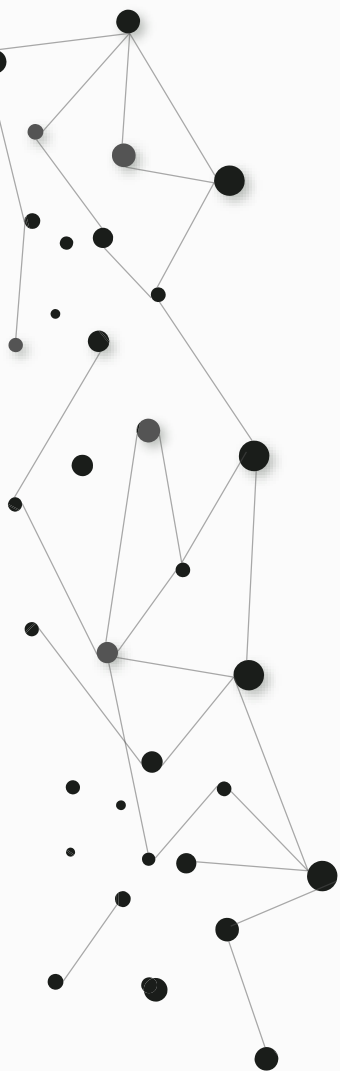
Part 02

Model

02/



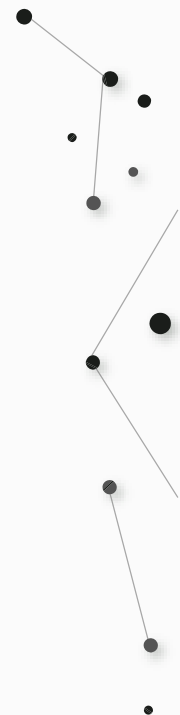
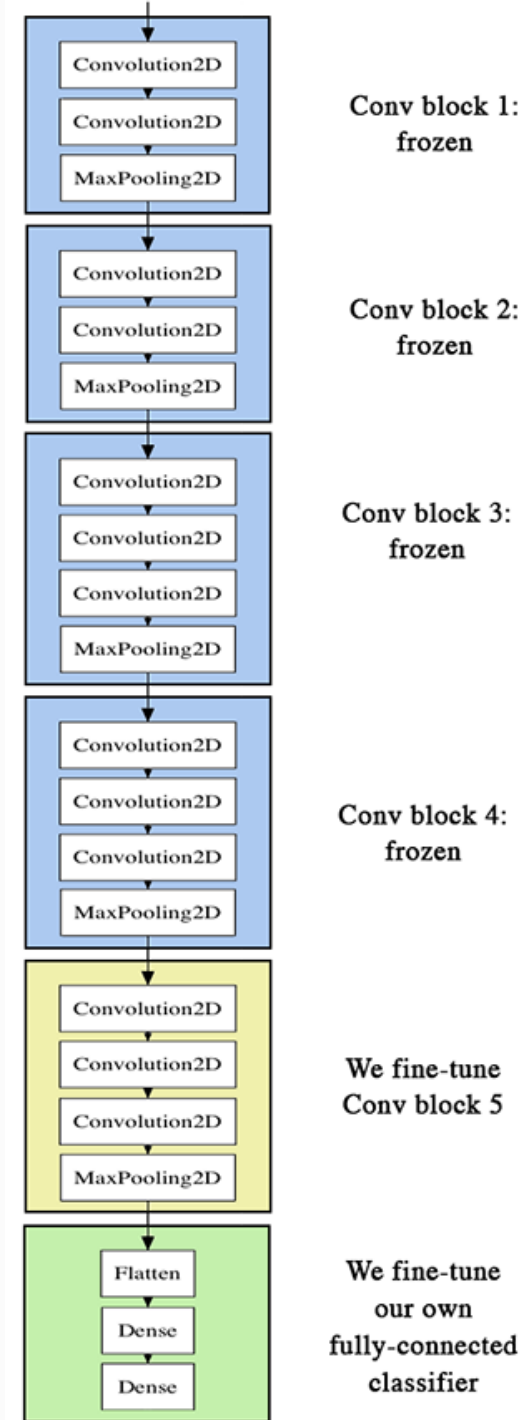
02/



● Main hallmark
Deeper is better,
Steadily increase the depth

● Conv
Small Conv is better

● LRN useless
Batch normalization will fail when
the Batch size is small



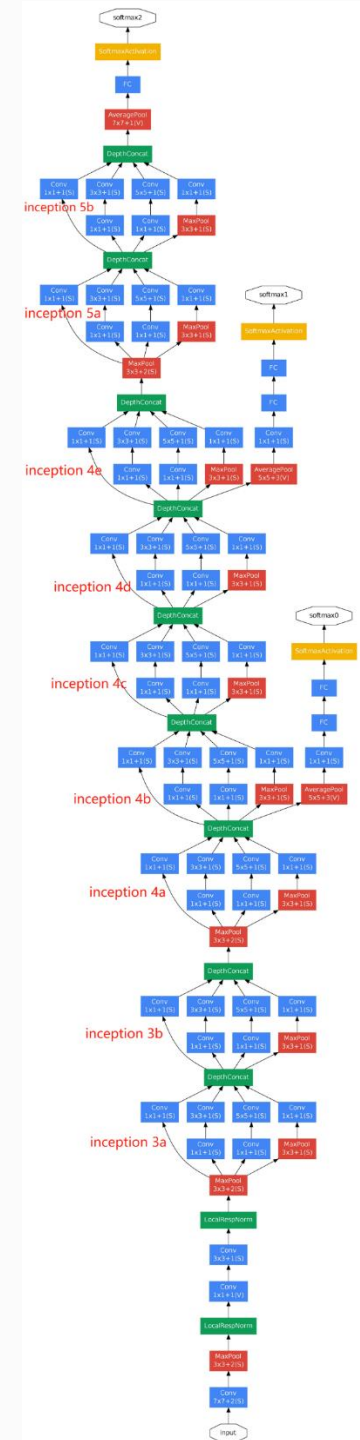
02/

GoogLeNet

Main hallmark
Improved utilization of the computing
resources inside the network

Deep network
22 Layers

Parameter
Uses 12 times fewer parameters



02/

ResNet

Main hallmark

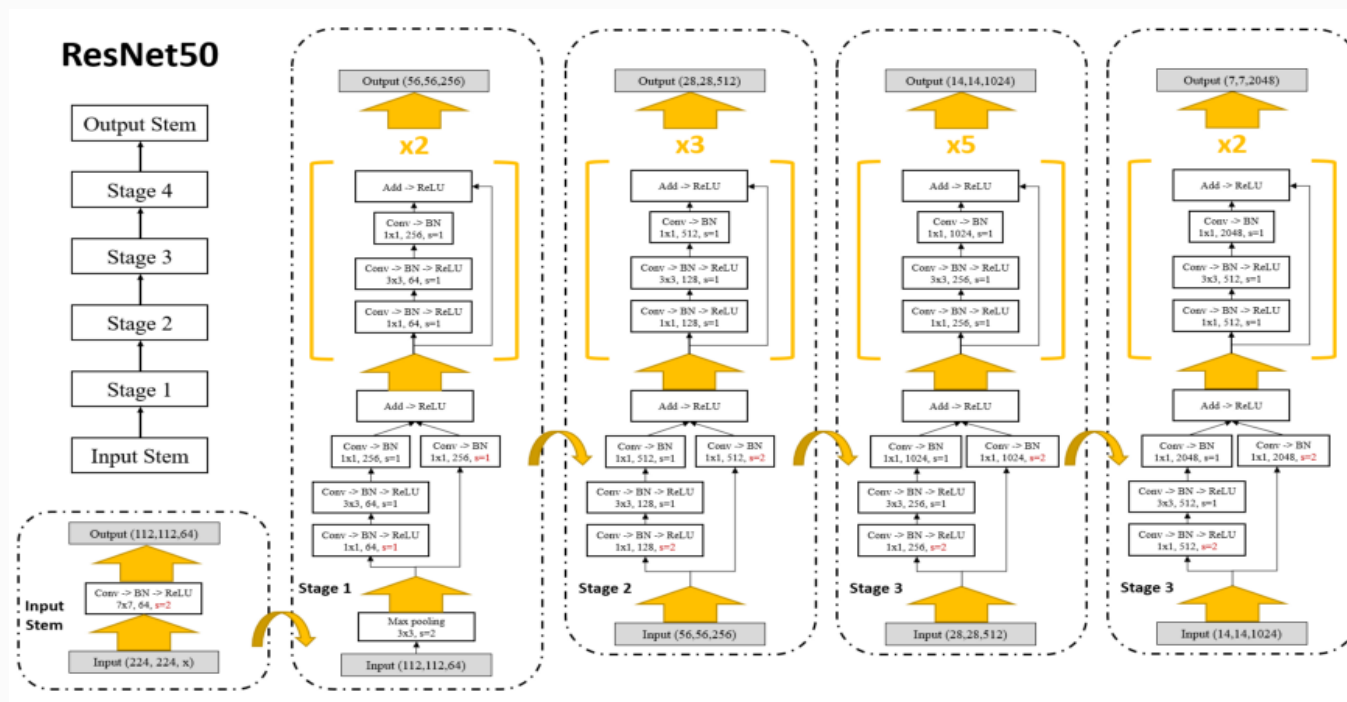
Solves the degradation problem
through residual learning

Deep network

152 Layers

Inconsistent maintenance

1. zero-padding
2. projection shortcut



DenseNet

Main hallmark

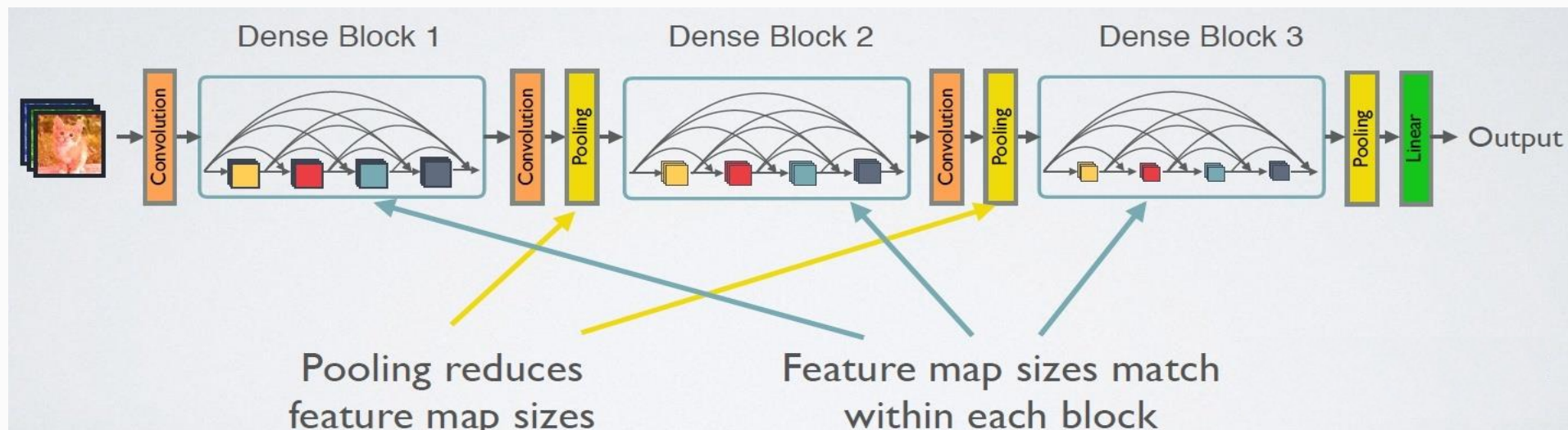
DenseBlock and Transition

Parameter

use of 1x1 convolutional layers reduces the dimensionality of feature maps

Prevention of Overfitting

Batch Normalization and Dropout are used to prevent overfitting and improve the model's generalization ability.



CoAtNet

Main hallmark

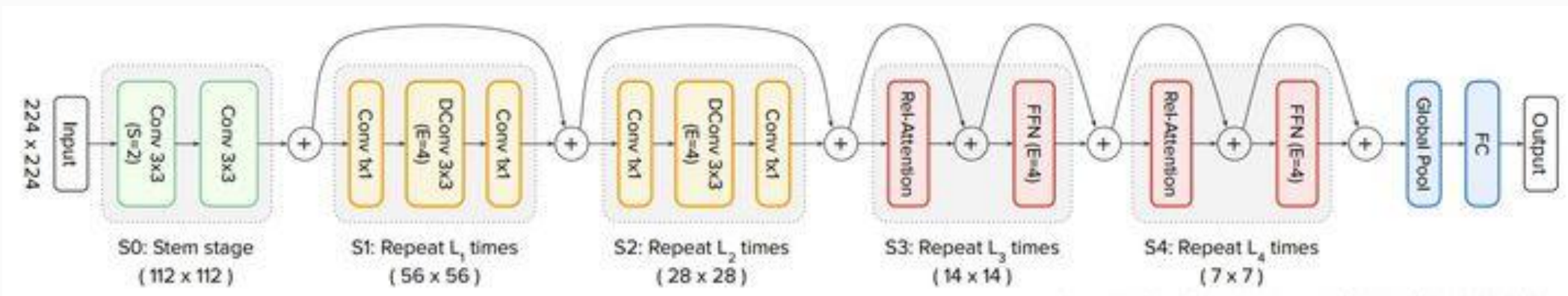
Merging Convolution and Self-Attention

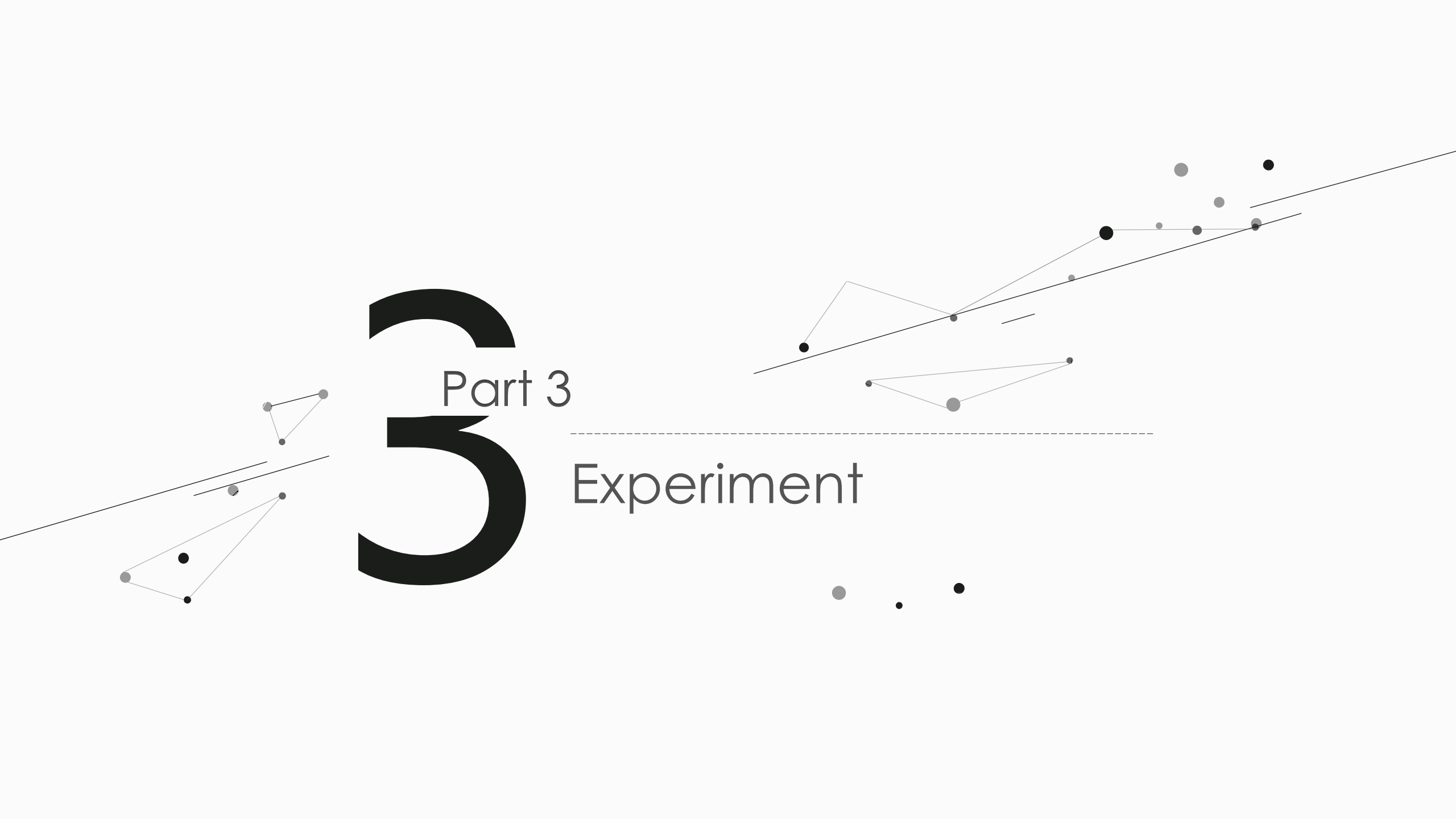
Lightweight Structure

reduces computation and storage costs by using shallow networks and downsampling

MBConv

1. Depthwise Convolution
2. inverted bottleneck





3

Part 3

Experiment

03

ResNet50



Hyperparameters

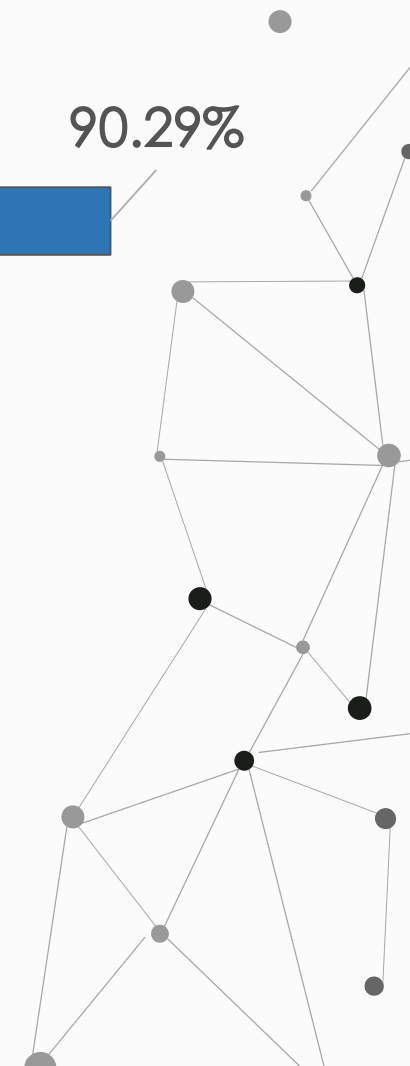
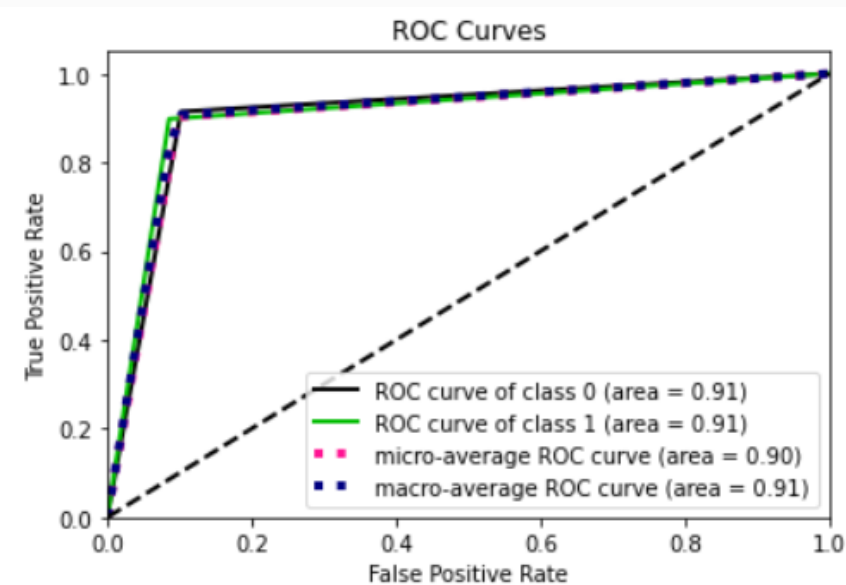
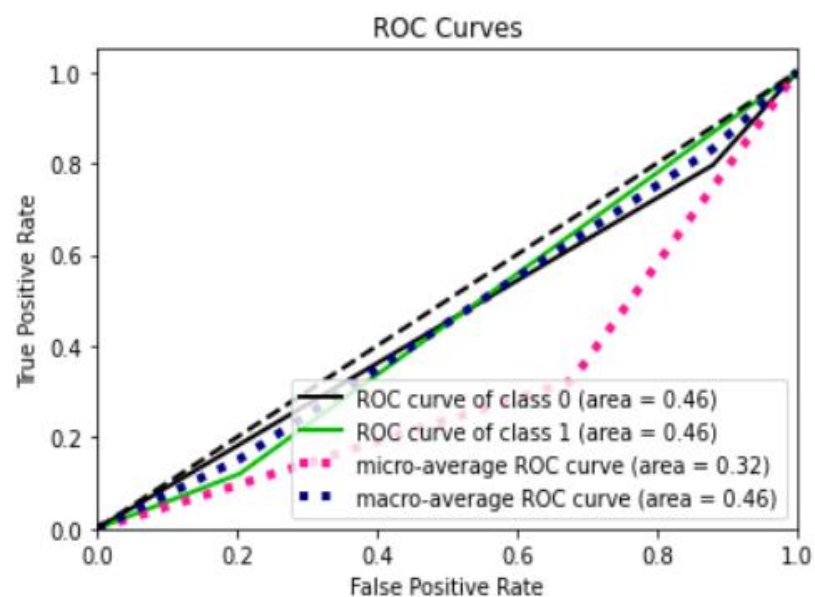
Optimizer

Adam

32.24%

SGD

90.29%



03

ResNet50



Hyperparameters

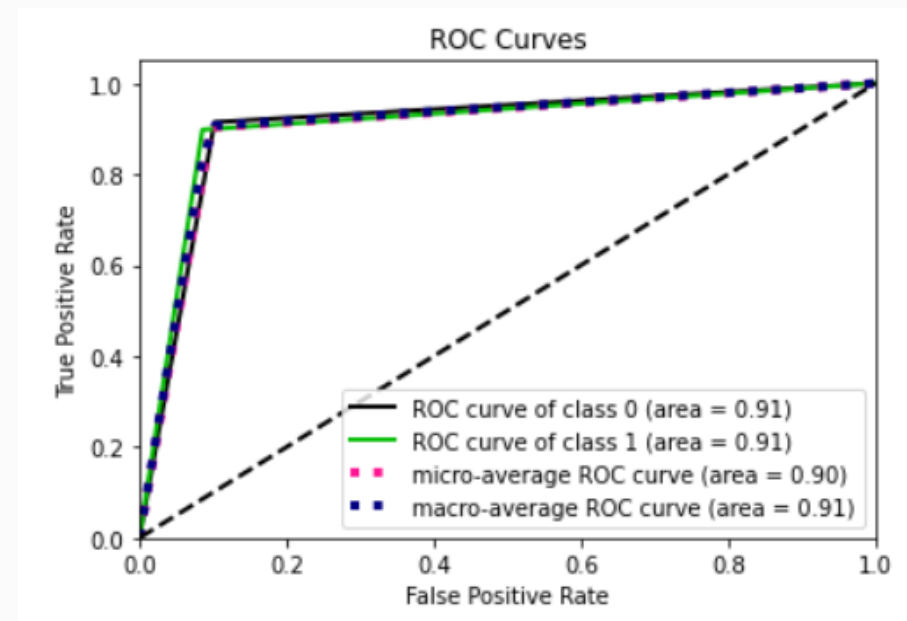
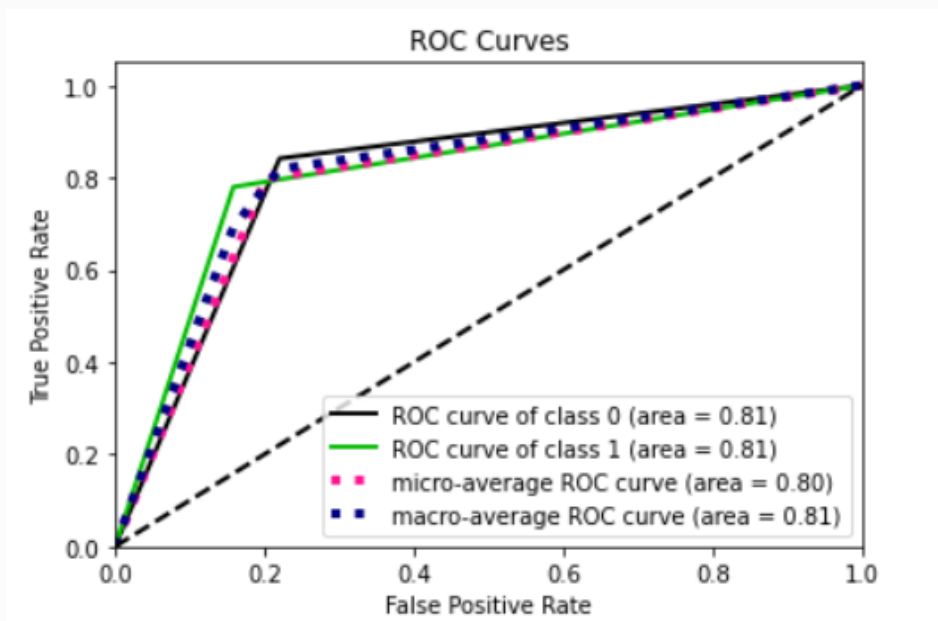
Learning rate

ConstLR

79.83%

StepLR

90.29%



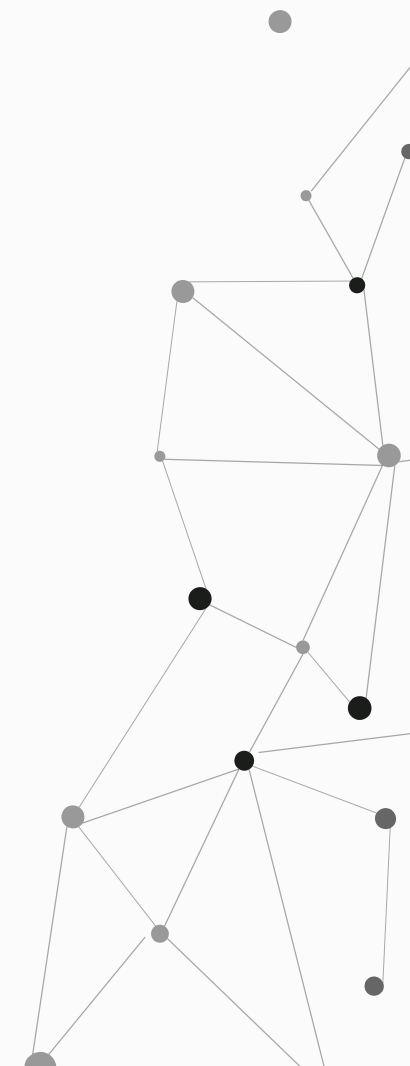
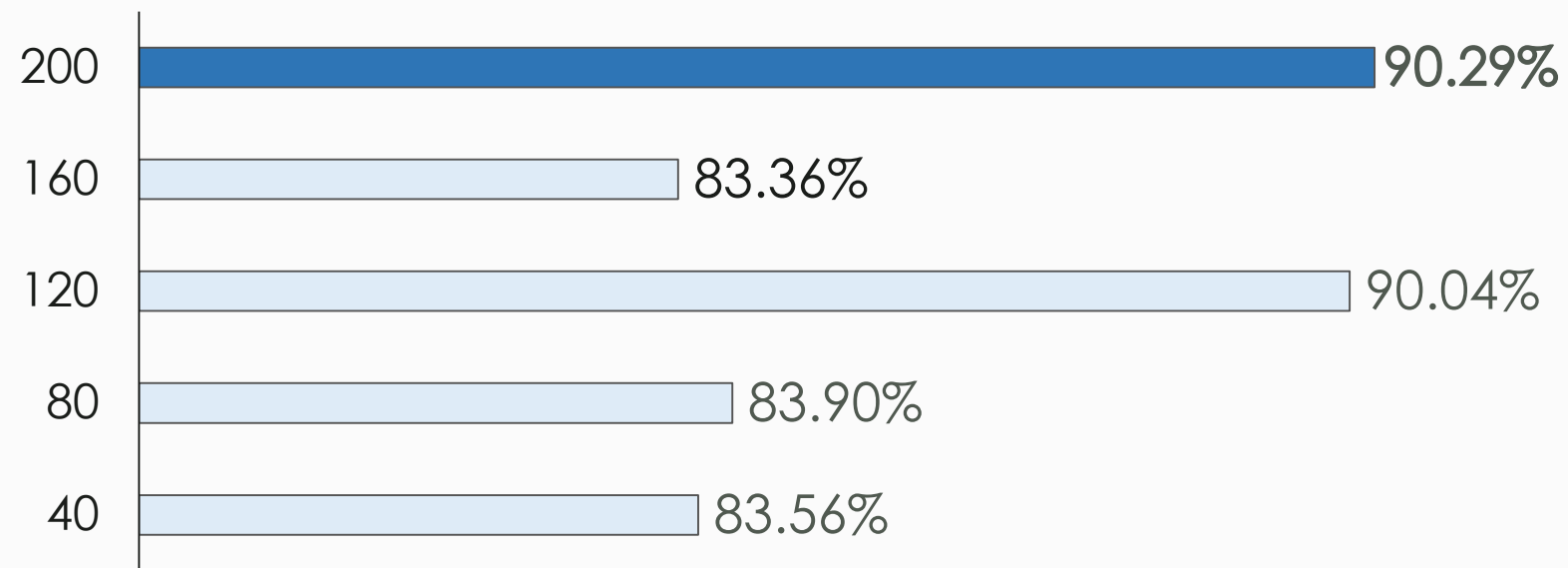
03/

ResNet50



Hyperparameters

Epochs

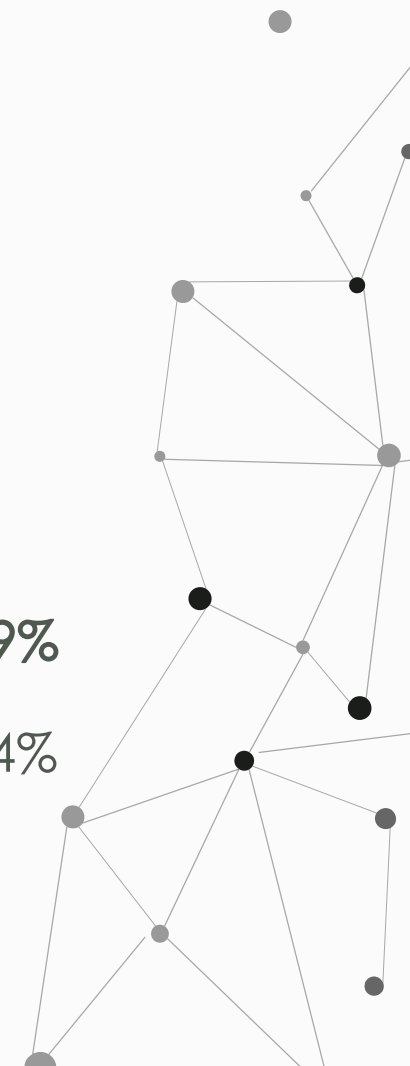
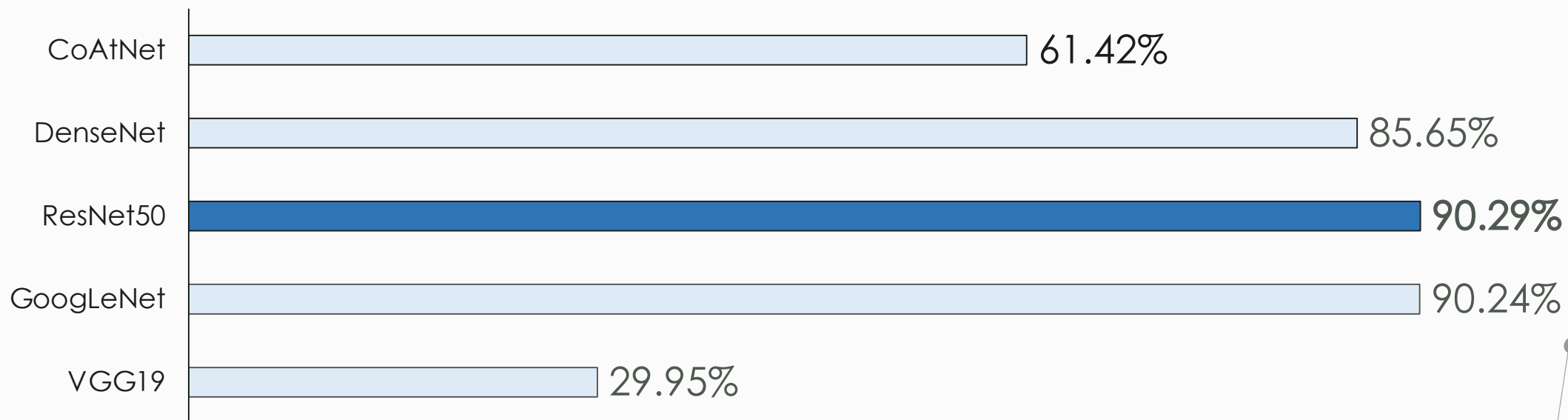


03/

Model



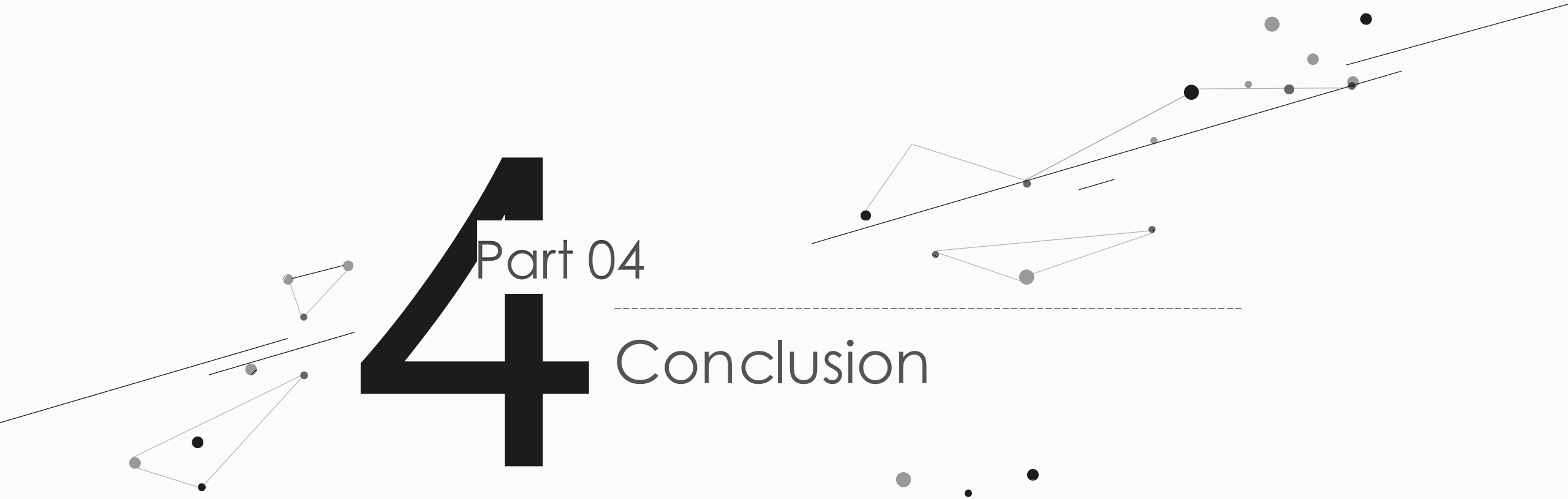
Test_Accuracy

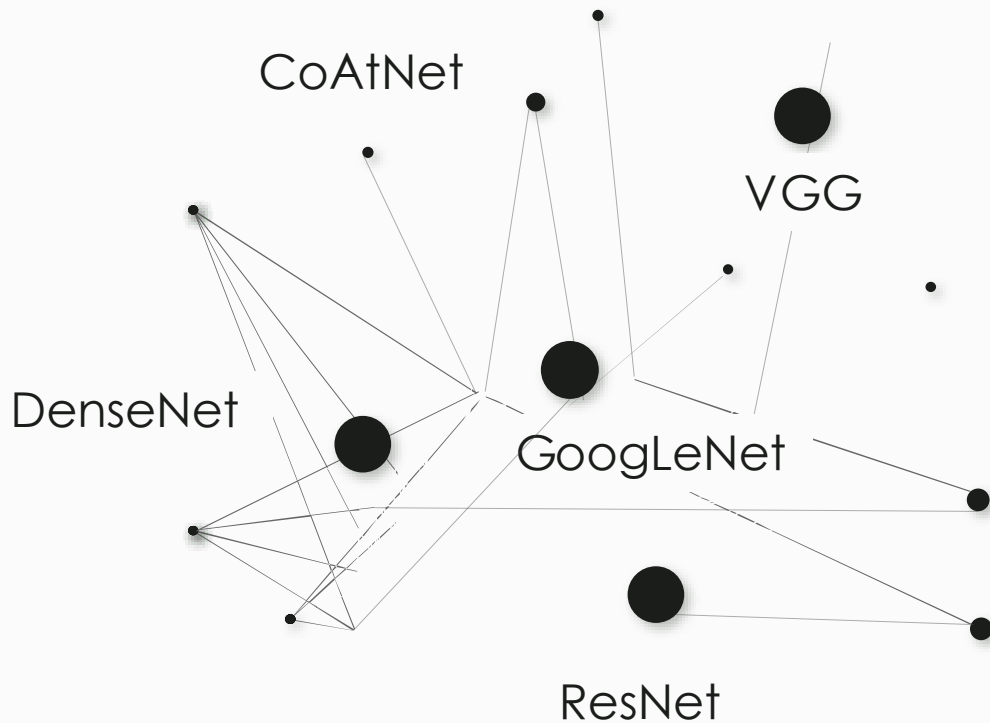


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Part 04

Conclusion





Conclusion

In this experiment, we study the effect of different models on this dataset.

We try to enhance the accuracy by using adjusting hyperparameters.

Constant learning rate vs step learning rate

Finally, we found that GoogLeNet and ResNet50 has a good balance between accuracy and generalization.



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Thank You