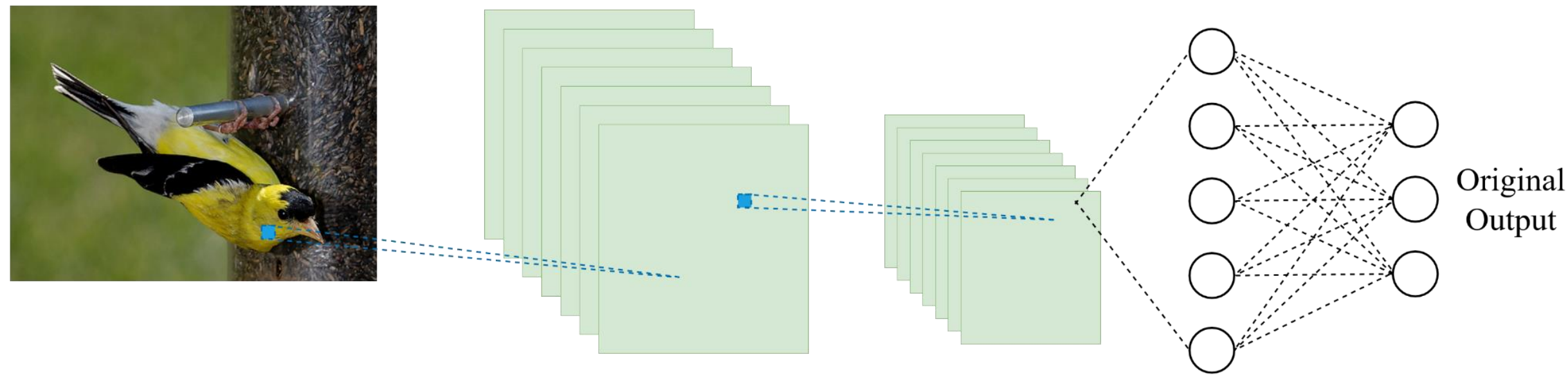


Tema:

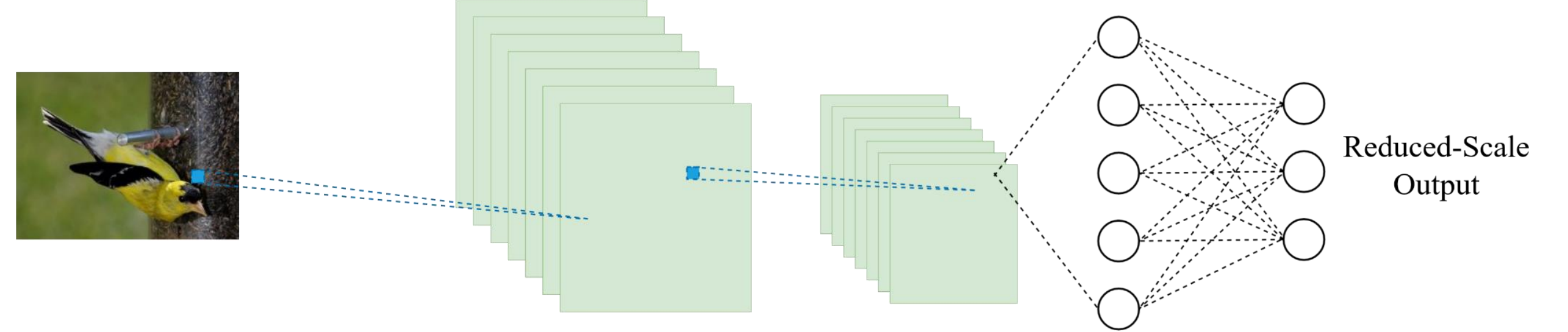
Resolution-wise Convolutional Neural Networks for Image Classification

Original Resolution (224x224)



Original Output

Reduced Resolution (196x196)



Reduced-Scale Output

Introduction

- Convolutional Neural Networks (CNNs) excel in computer vision tasks, especially image classification
- Modern CNN models and higher image resolutions increase computational costs, affecting portability and resource requirements
- In this work we explore how resolution affects CNNs, balancing between accuracy and increased memory and processing needs
- Our goal is to minimize network parameters, enhancing suitability for resource-constrained devices such as mobile phones.

Problem Statement

- Research Questions**
 - Is it possible to reduce the resolution of a trained Neural Network model and keep its predictive capacity?
 - Is it possible to systematically encounter a resolution for which the model keeps its predictive ability?
- Answering these questions enables us to reduce the computational cost of Convolutional Neural Network-based solutions by simply reducing the resolution of pre-trained models

Experiments

- We use two well-known benchmarks, CIFAR-10 and ImageNet, and popular CNN architectures: NasNet, ResNet, and MobileNet
- Main Experiments**
 - Reduce the resolution of pre-trained CNNs and evaluate the floating points operations (FLOPs) and accuracy compared to the original model
 - Select the resolution randomly from a set of pre-defined resolutions, and evaluate the FLOPs and accuracy compared to the original model

Results

- Resolution Reduction
 - NasNet on CIFAR-10

| Resolution | FLOPs Drop (%) | Accuracy Drop (%) |
|------------|----------------|-------------------|
| 28 x 28 | 23.43 | 1.92 |
| 24 x 24 | 43.74 | 4.34 |
| 20 x 20 | 60.93 | 13.32 |
| 16 x 16 | 74.99 | 29.58 |
| 12 x 12 | 85.93 | 52.40 |

- ResNet on ImageNet: Comparison with existing works

| Resolution | FLOPs Drop (%) | Accuracy Drop (%) |
|--------------------------|----------------|-------------------|
| He et al. [1] | 20.00 | 1.70 |
| Wang et al. [2] | 20.00 | 2.00 |
| He et al. [3] | 20.00 | 1.40 |
| Ours (210 x 210) | 3.90 | 1.22 |
| Ours (196 x 196) | 14.72 | 2.90 |
| Ours (182 x 182) | 29.26 | 2.70 |
| Ours (168 x 168) | 38.49 | 5.13 |
| Ours (154 x 154) | 49.50 | 6.57 |
| Ours (140 x 140) | 57.26 | 10.72 |
| Ours (Random Resolution) | 17.30 | 2.42 |

Conclusions

- The experiments performed on ImageNet dataset suggest that **it is possible to reduce the CNN resolution with minor reduction on its predictive capability**
- These experiments achieved **significant reductions in FLOPs with a minor drop in accuracy**
- Random selection of model resolutions in experiments showed FLOPs reduction comparable to other methods in literature, with minimal accuracy loss
- There is potential for developing a more refined method for selecting the scale of Neural Networks

References

- [1] HE, Y.; ZHANG, X.; SUN, J. Channel pruning for accelerating very deep neural networks. In: IEEE international conference on computer vision. [S.l.: s.n.], 2017
- [2] WANG, X. et al. Skipnet: Learning dynamic routing in convolutional networks. In: European Conference on Computer Vision (ECCV) [S.l.: s.n.], 2018
- [3] HE, Y. et al. Amc: Automl for model compression and acceleration on mobile devices. In: European Conference on Computer Vision (ECCV) [S.l.: s.n.], 2018.