

# Dog Emotion Prediction Based On Transfer Learning

Team: Tree New Bee

Wei Shan, Jeff Zhuo, Xi Du, Tianrui Ye, Yuewei Wang



# Introduction

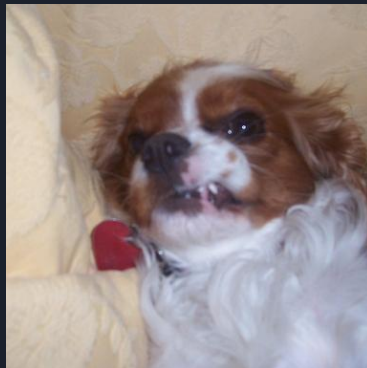
- Motivation
  - We love Dogs
- Objective
  - Image Classification
    - 4 classes
- Background
  - Transfer Learning



# Introduction

- Motivation
  - We love Dogs
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  - Emotion Classification
    - 4 classes
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Angry



Happy



Relaxed

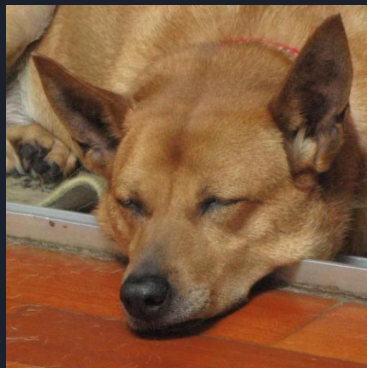
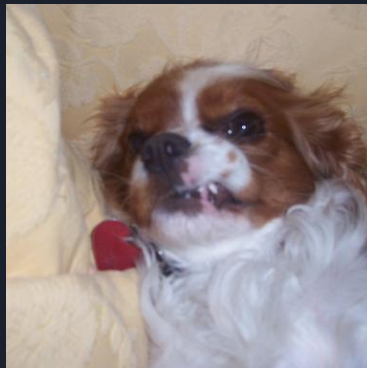


Sad

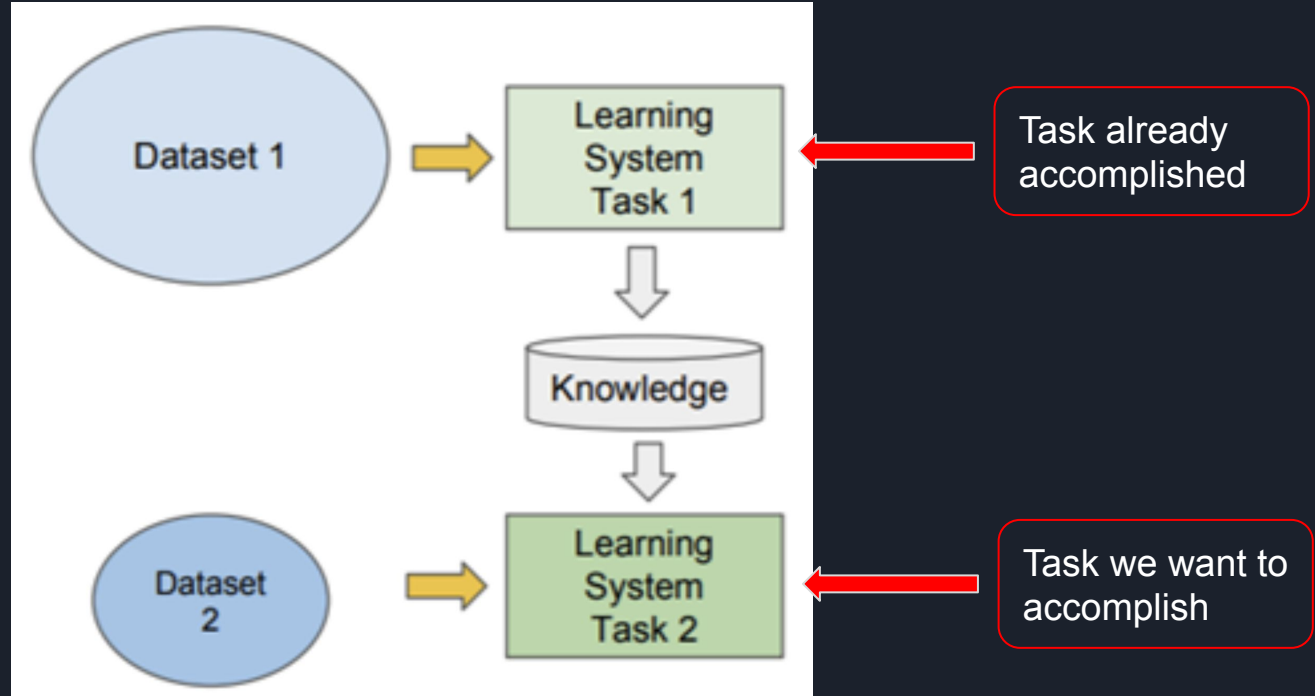


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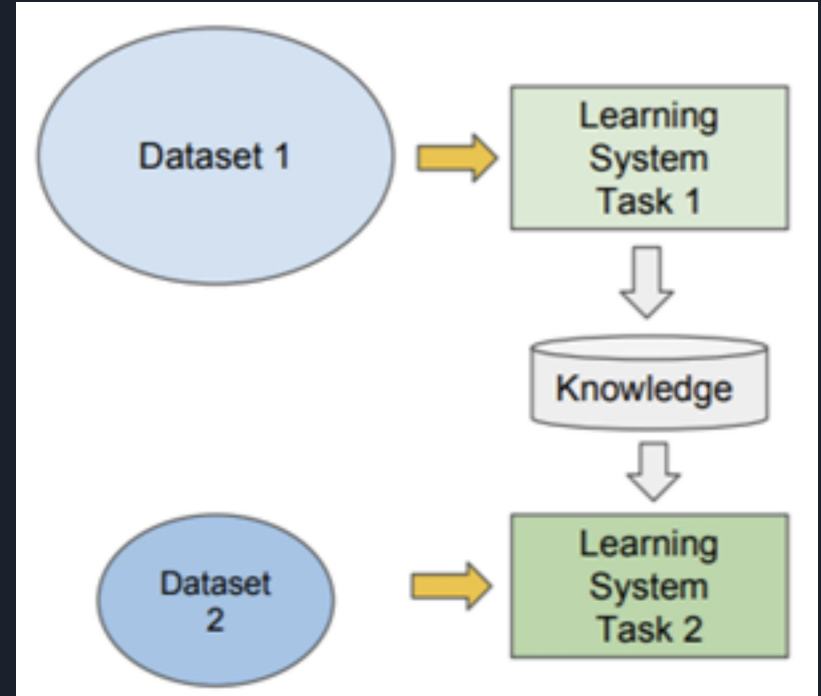


# Introduction(Background - Transfer Learning)



# Introduction(Background - Transfer Learning)

- Task A and B have similar input  $x$ .
- You have a lot more data for Task A than Task B.
- Low level features from A could be helpful for learning B.

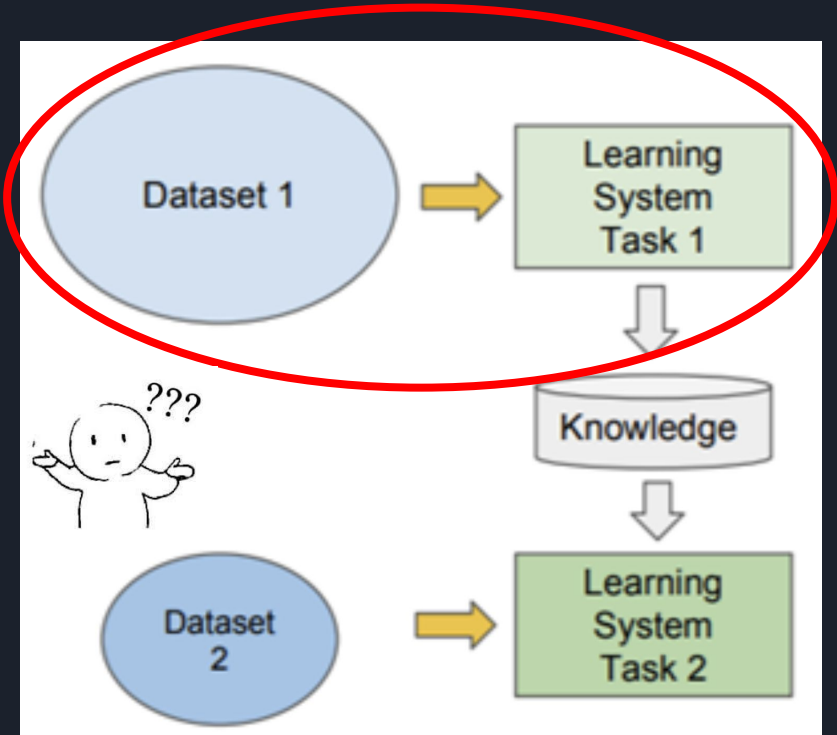


Works cited: Dr. Andrew Ng, <https://www.youtube.com/watch?v=yofjFQddwHE>

Image Credit: Prof. Li

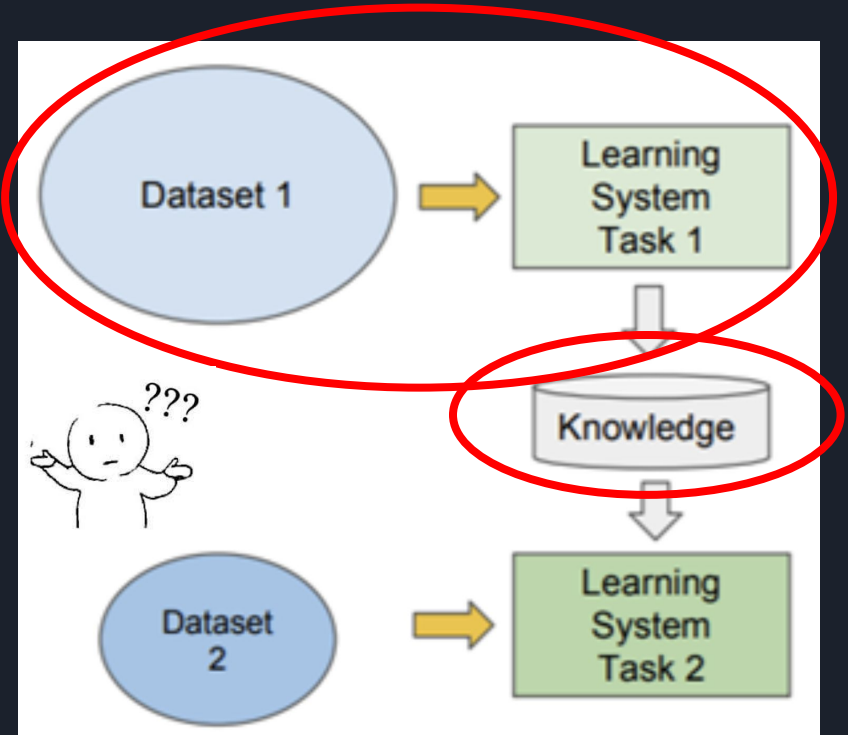
# Introduction(Background - Transfer Learning)

- Model trained by big dataset?



# Introduction(Background - Transfer Learning)

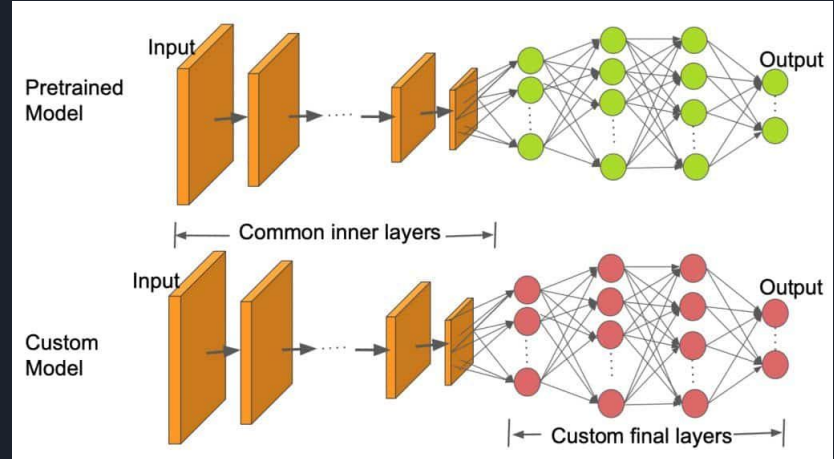
- Model trained by big dataset?
- How to transfer knowledge?





# Introduction(Background - Transfer Learning)

Q: Model trained by big dataset?

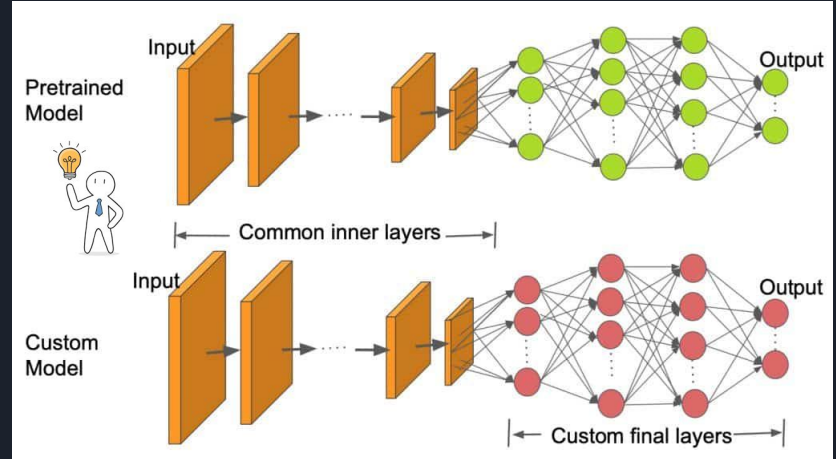


# Introduction(Background - Transfer Learning)

Q: Model trained by big dataset?

A: Pre-trained Model!

- ResNet
- VGG
- AlexNet
- MobileNet
- GoogLeNet



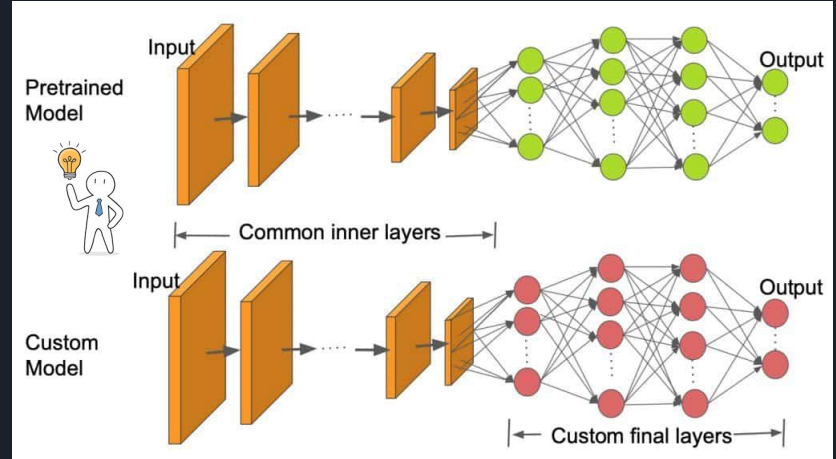
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Q: How to transfer knowledge?



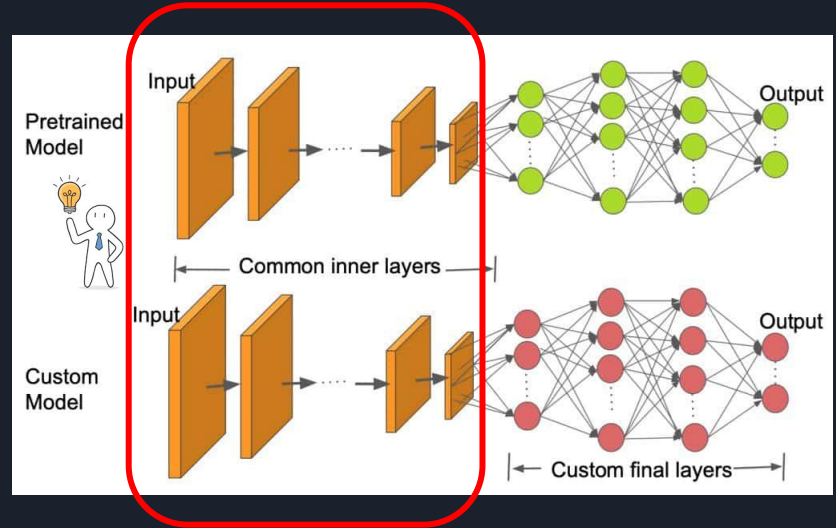
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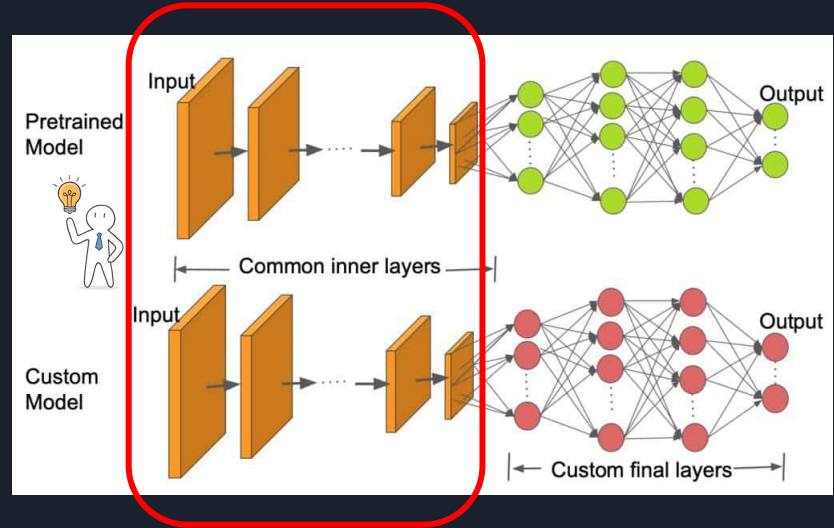
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Q: How to transfer knowledge?

A: Freeze convolutional layers' parameters!



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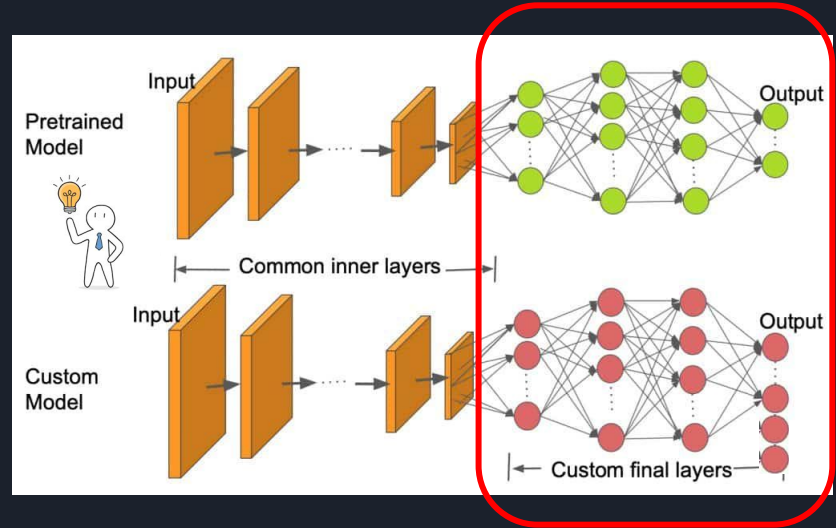
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A: Freeze convolutional layers' parameters!





# Review of Literature



# Review of Literature: Image-based Sentiment Analysis

- Lei Zhang, Shuai Wang, & Bing Liu. (2018). Deep Learning for Sentiment Analysis : A Survey.
- Mittal, N., Sharma, D., & Joshi, M. (2018). Image Sentiment Analysis Using Deep Learning. In 2018 IEEE/WIC/ACM International Conference on Web Intelligence (WI) (pp. 684-687).
- Cetinic, E., Lipic, T., & Grgic, S. (2019). A Deep Learning Perspective on Beauty, Sentiment, and Remembrance of Art. IEEE Access, 7, 73694-73710.





# Review of Literature: Pre-trained Models

- Kaiming He, Xiangyu Zhang, Shaoqing Ren, & Jian Sun. (2015). Deep Residual Learning for Image Recognition.
- Karen Simonyan, & Andrew Zisserman. (2015). Very Deep Convolutional Networks for Large-Scale Image Recognition.
- Krizhevsky, A., Sutskever, I. & Hinton, G. E. (2012). ImageNet Classification with Deep Convolutional Neural Networks.
- Andrew G. Howard, Menglong Zhu, Bo Chen, Dmitry Kalenichenko, Weijun Wang, Tobias Weyand, Marco Andreetto, & Hartwig Adam. (2017). MobileNets: Efficient Convolutional Neural Networks for Mobile Vision Applications.
- Christian Szegedy, Wei Liu, Yangqing Jia, Pierre Sermanet, Scott Reed, Dragomir Anguelov, Dumitru Erhan, Vincent Vanhoucke, & Andrew Rabinovich. (2014). Going Deeper with Convolutions.



# Review of Literature: Transfer Learning

- Pan, S., Tsang, I., Kwok, J., & Yang, Q. (2011). Domain Adaptation via Transfer Component Analysis. *IEEE Transactions on Neural Networks*, 22(2), 199-210.
- Jason Yosinski, Jeff Clune, Yoshua Bengio, & Hod Lipson. (2014). How transferable are features in deep neural networks?
- Y. Zhang and Q. Yang. A Survey on Multi-Task Learning.

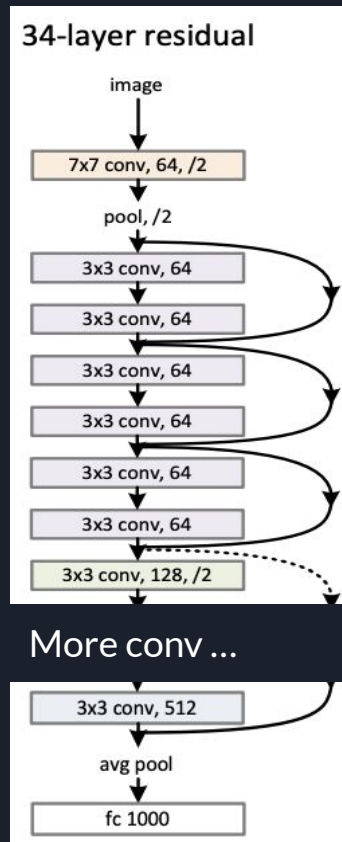


# Methodology (Pre-trained Model Selection)

- Pre-trained Model
  - ResNet
  - VGG
  - AlexNet
  - MobileNet
  - GoogLeNet

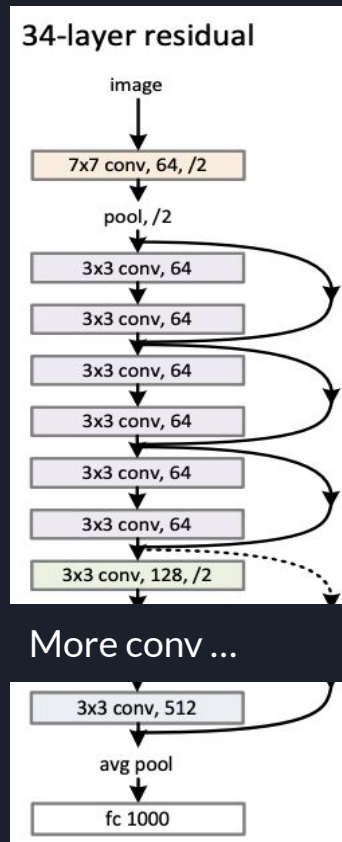
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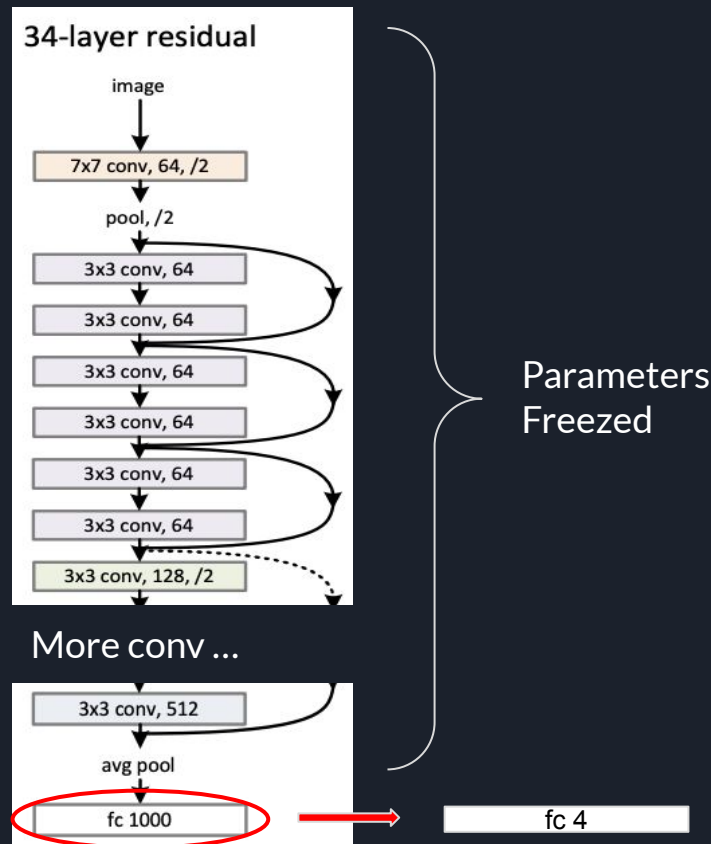
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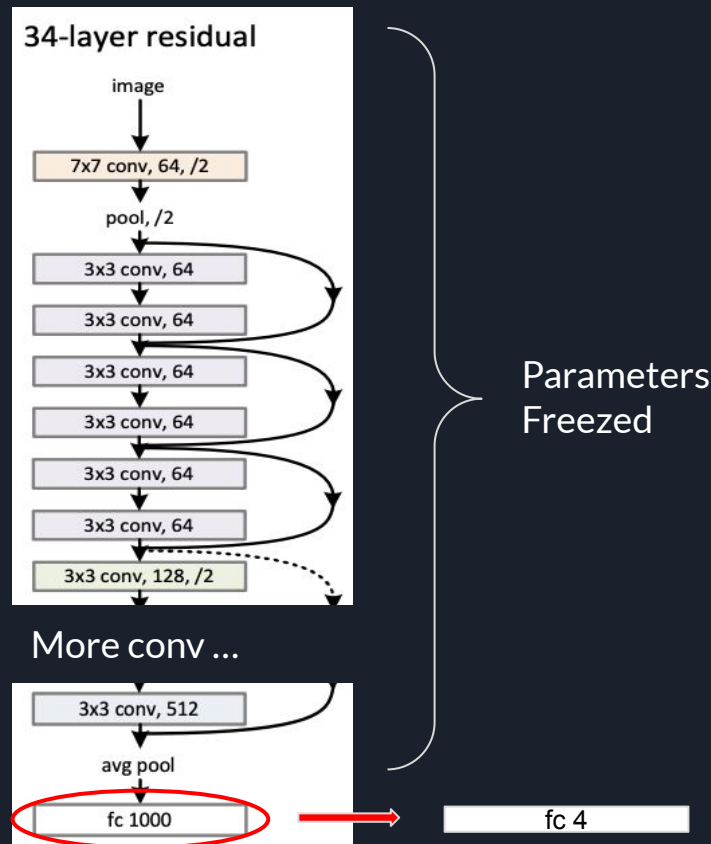
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- Change the last fully connected (fc) layer
  - Output size: 1000 -> 4



# Methodology (Pre-trained Model Selection)

- Pre-trained Model
  - ResNet
  - VGG
  - AlexNet
  - MobileNet
  - GoogLeNet
- Change the last fully connected (fc) layer
  - Output size: 1000 -> 4
- Accuracy Comparison





## Result (Pre-trained Model Selection)

Model	ResNet18	ResNet50	VGG16	AlexNet	MobileNet	GoogLeNet
Training Accuracy	44%	54%	58%	47%	51%	46%



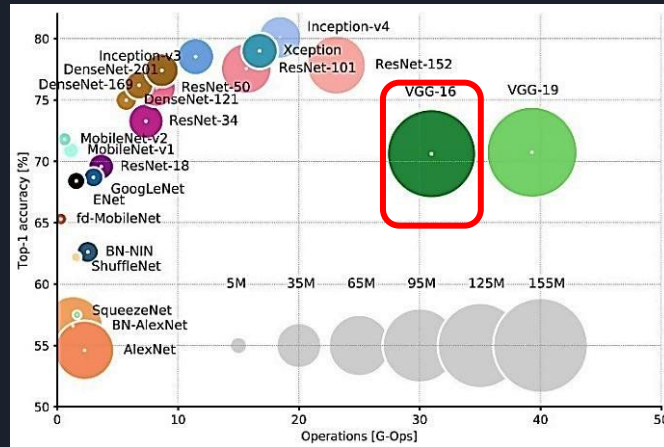


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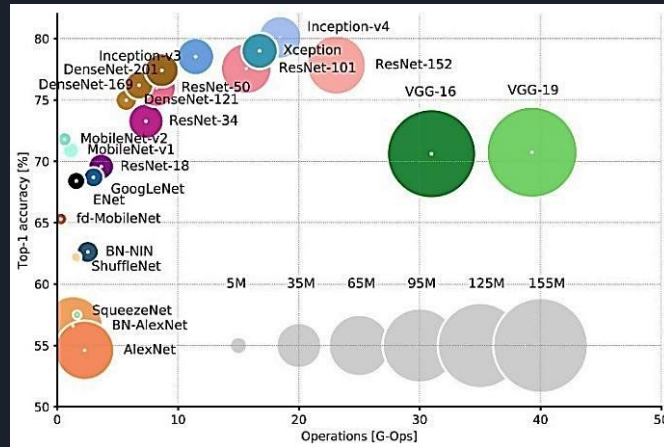
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# Methodology

- Classifier Modification
- Multi Task Learning
- Fine-tuning
- Unsupervised Domain Adaptation

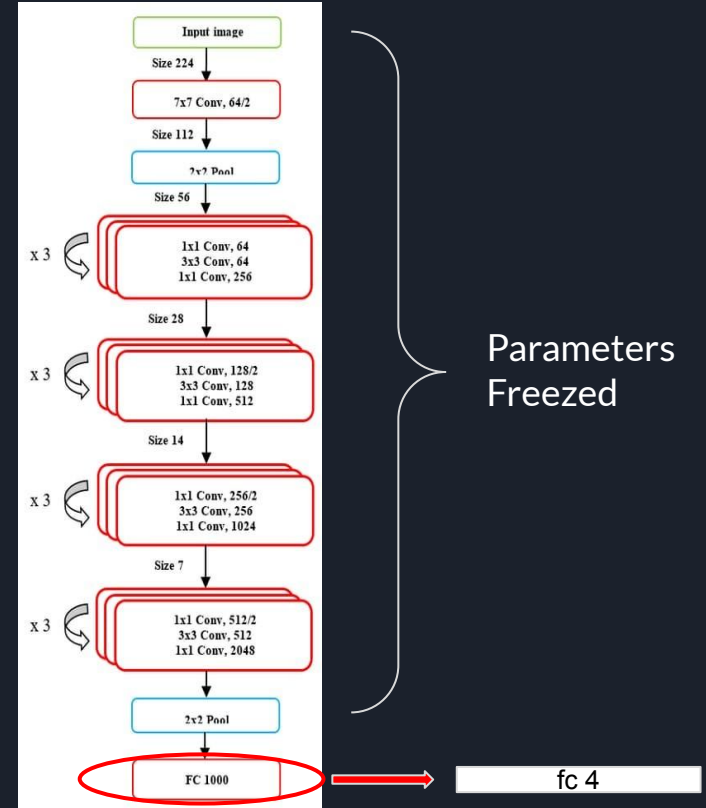


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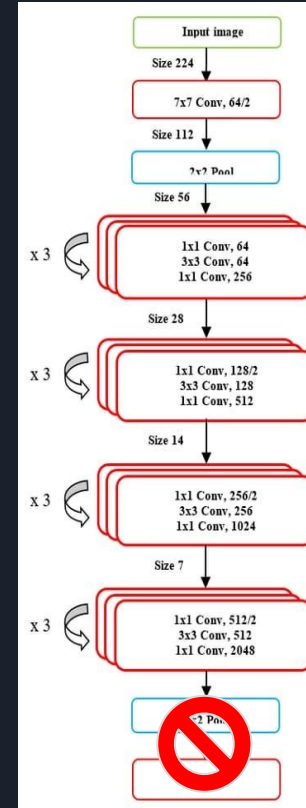
# Methodology (Classifier Modification)

- Fully Connected Layer



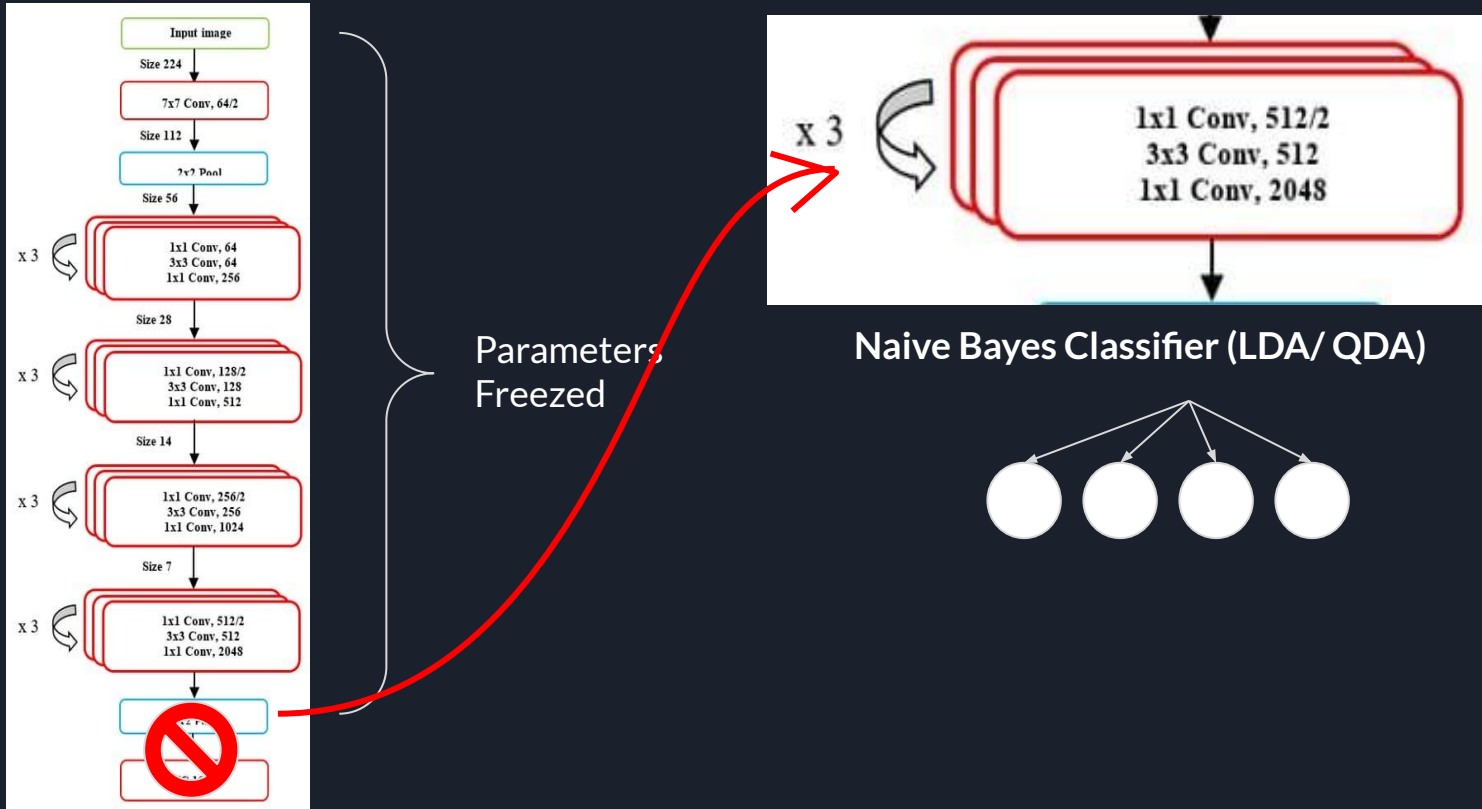
# Methodology (Classifier Modification)

- Fully Connected Layer
- Naive Bayes Classifier (LDA/ QDA)
- KNN
- SVM
- Random Forest



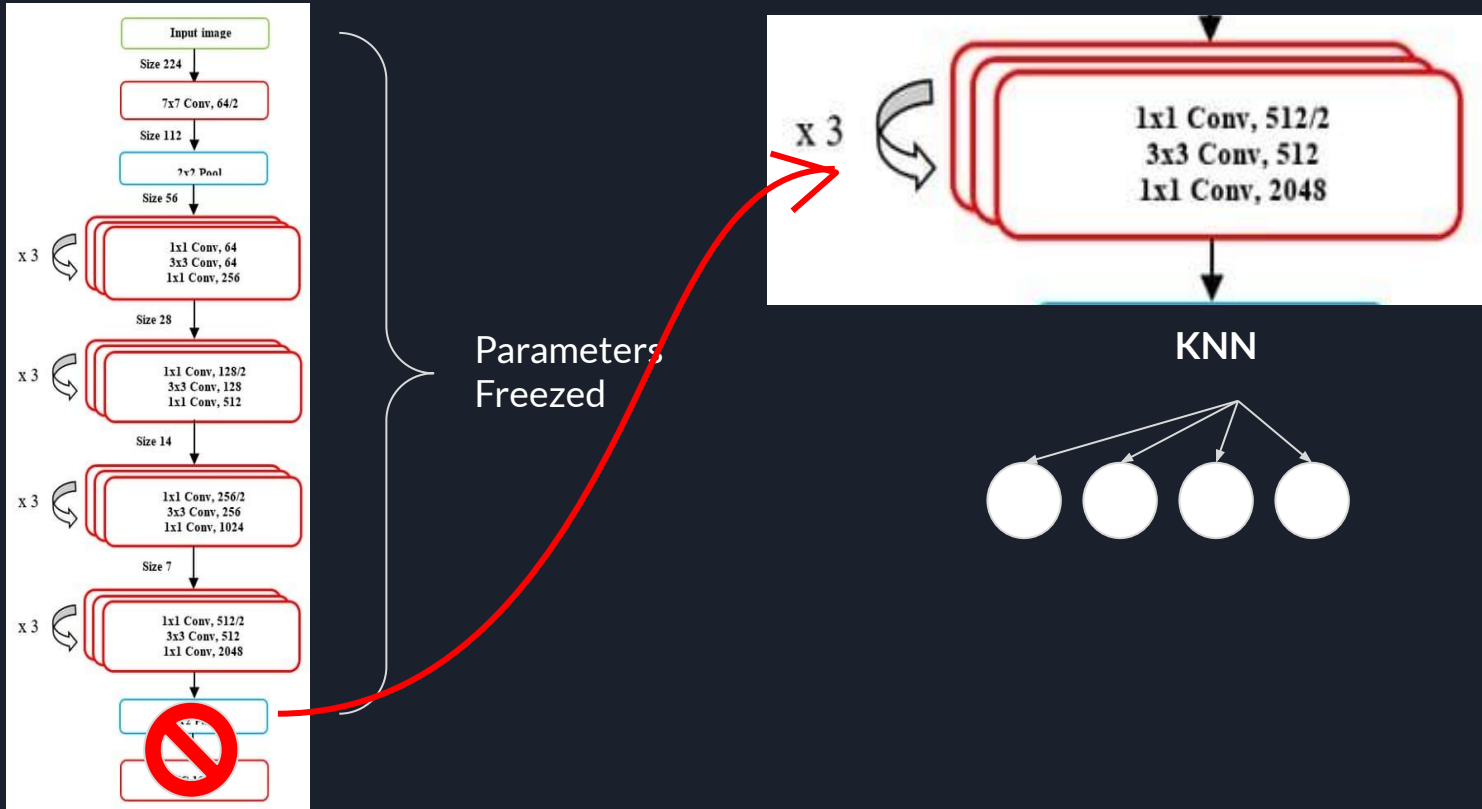
Parameters  
Frozen

# Methodology (Classifier Modification)

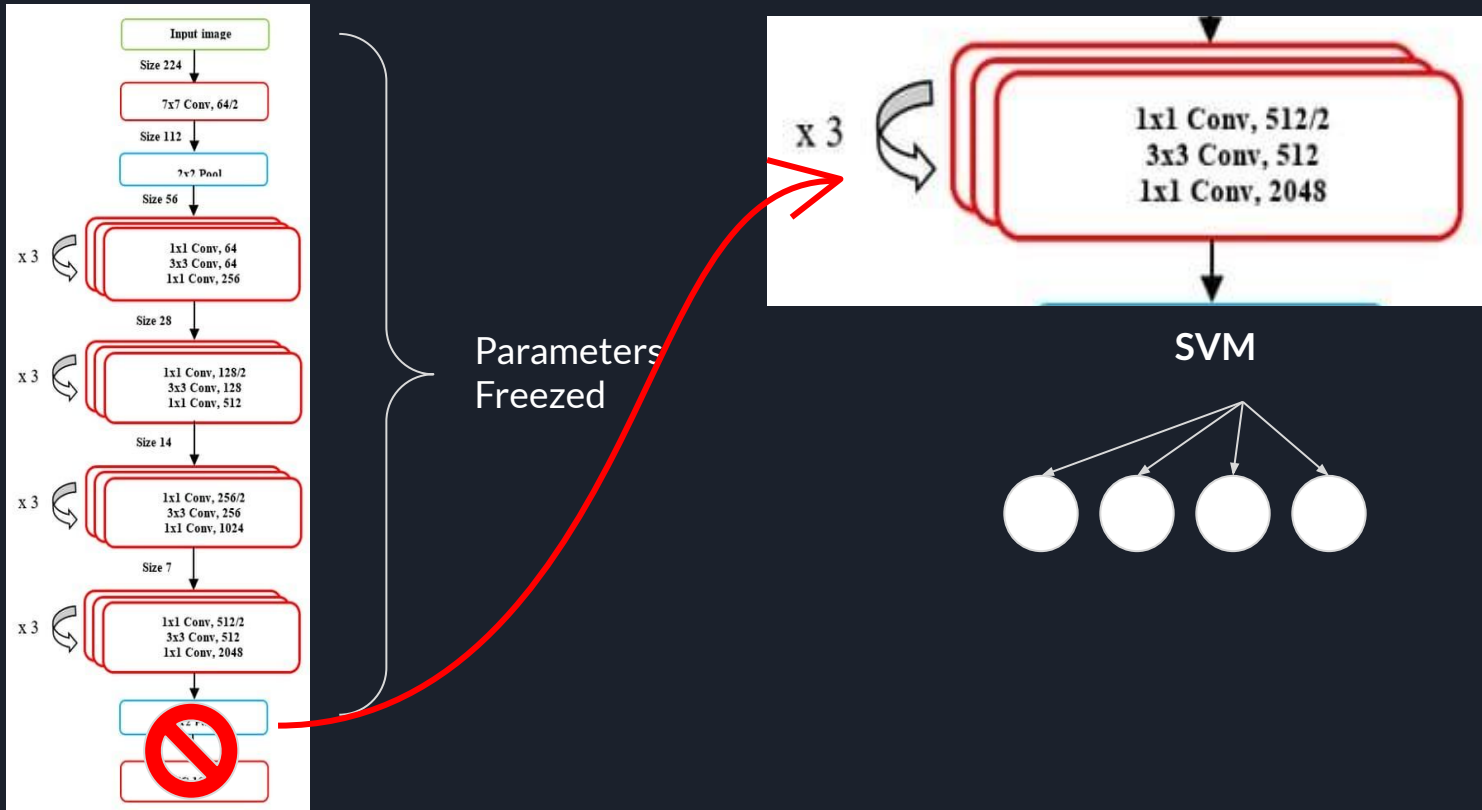




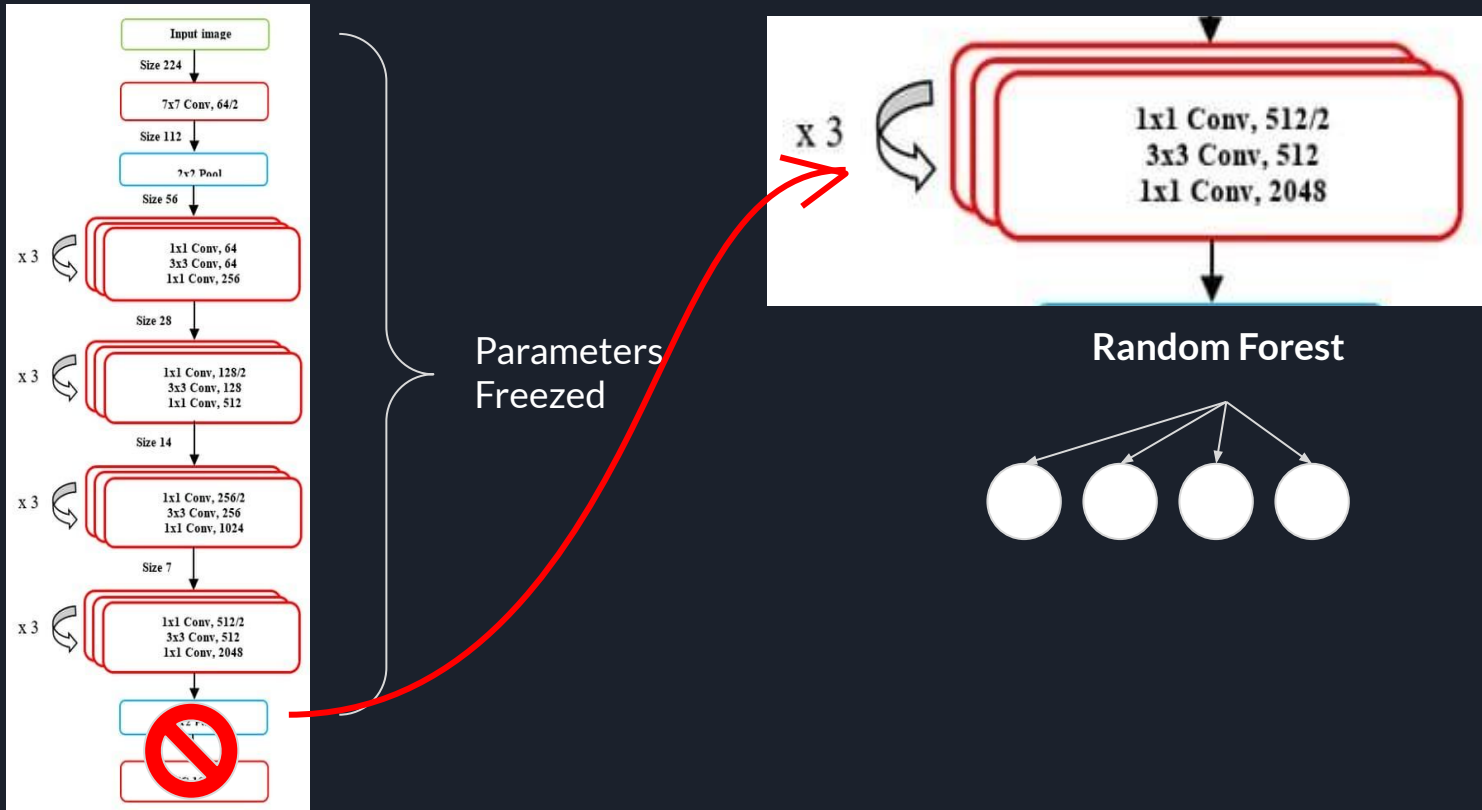
# Methodology (Classifier Modification)



# Methodology (Classifier Modification)



# Methodology (Classifier Modification)





## Result (Classifier Modification)

Model	FC	LDA	QDA	KNN	SVM	Random Forest
Training Accuracy	54%	79%	82%	48%	79%	61%



## Result (Classifier Modification)

Model	FC	LDA	QDA	KNN	SVM	Random Forest
Training Accuracy	54%	79%	82%	48%	79%	61%
Testing Accuracy	53%	31%	35%	20%	59%	27%

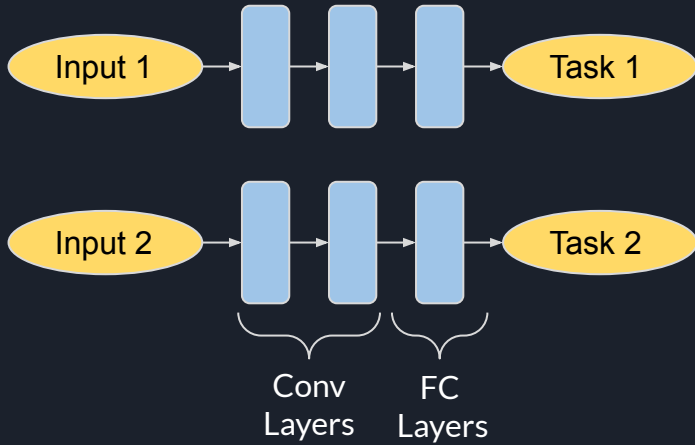


# Methodology

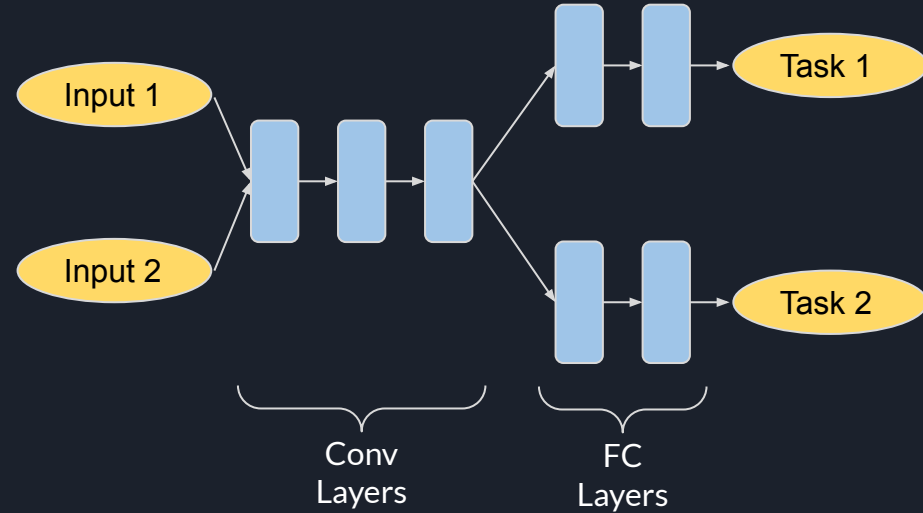
- Classifier Modification
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# Methodology (Multi Task Learning)

Single Task Learning

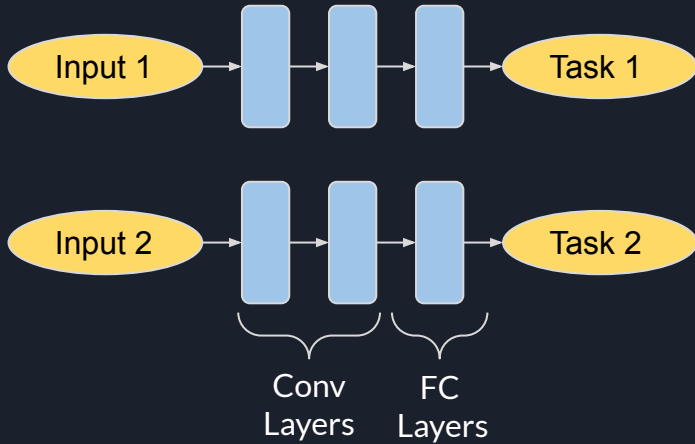


Multi Task Learning



# Methodology (Multi Task Learning)

Single Task Learning



- Tasks should be related.

Multi Task Learning

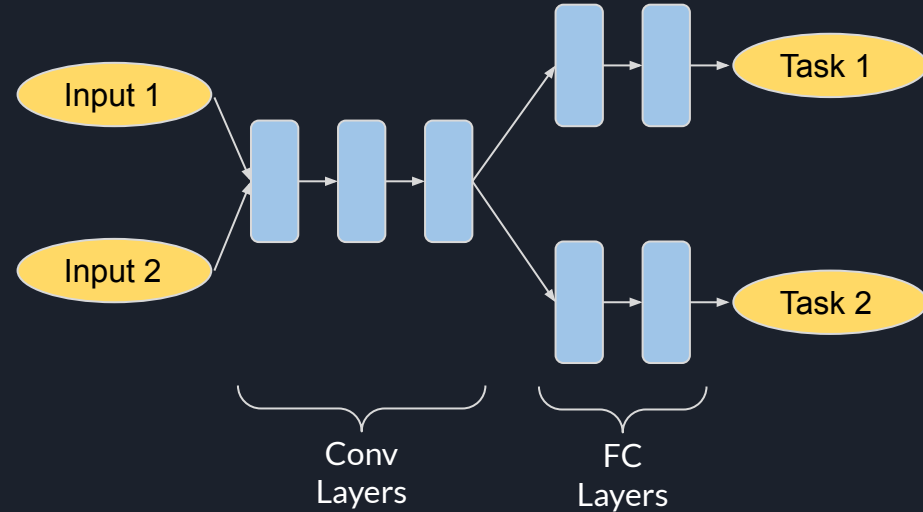
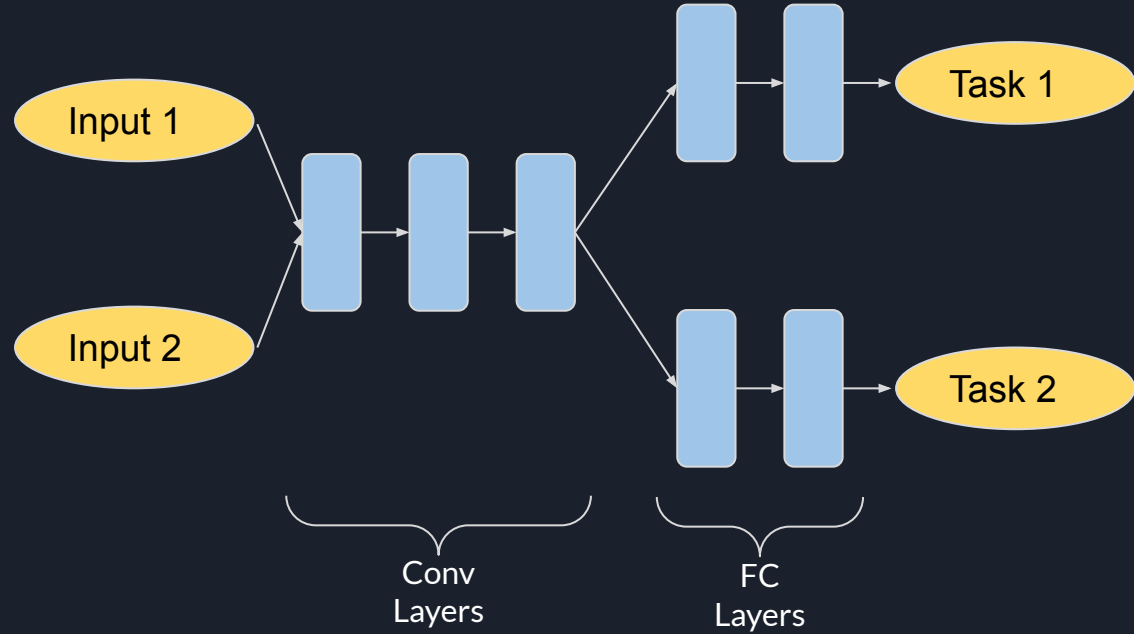


Image Credit: I made them

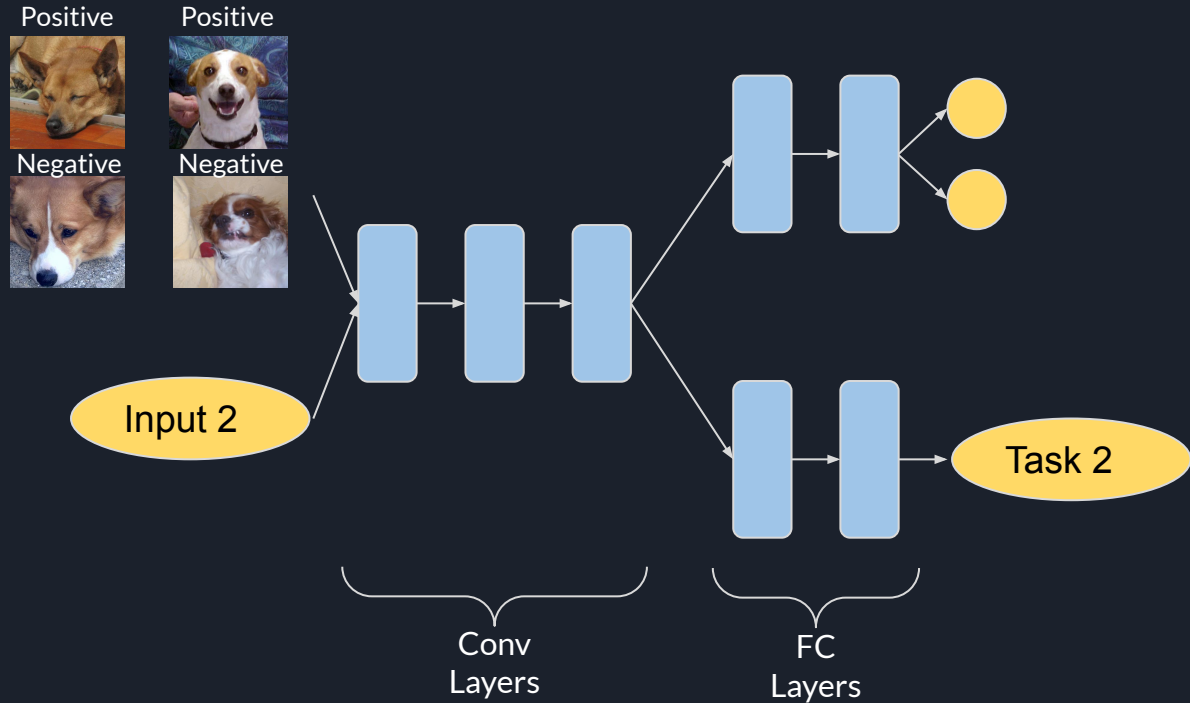
Works Cited: Y. Zhang and Q. Yang, "A Survey on Multi-Task Learning,"



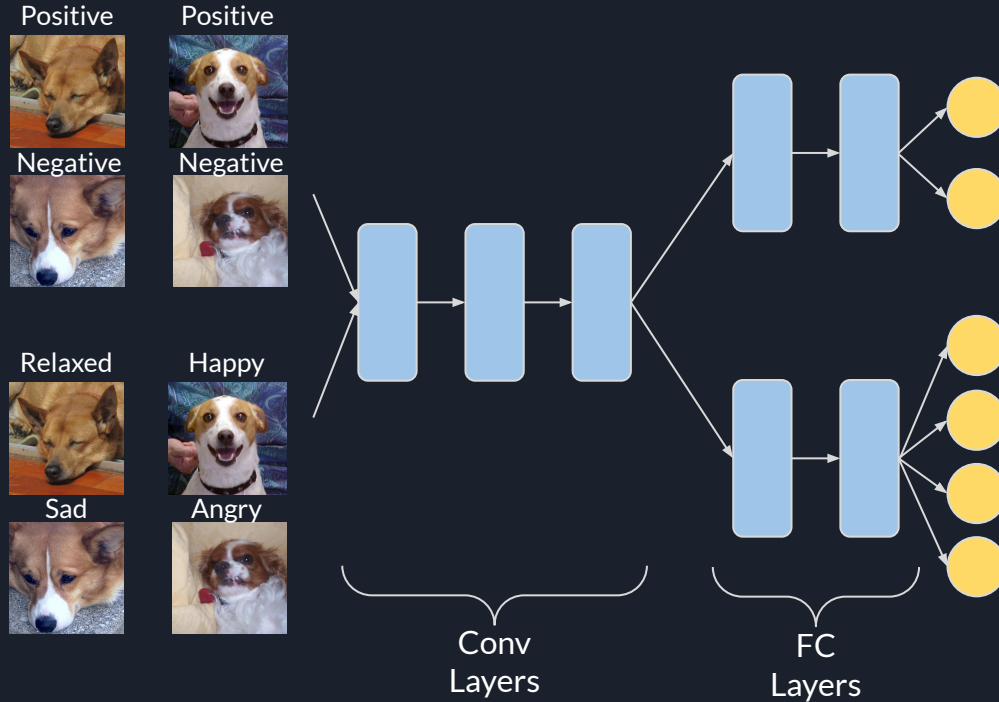
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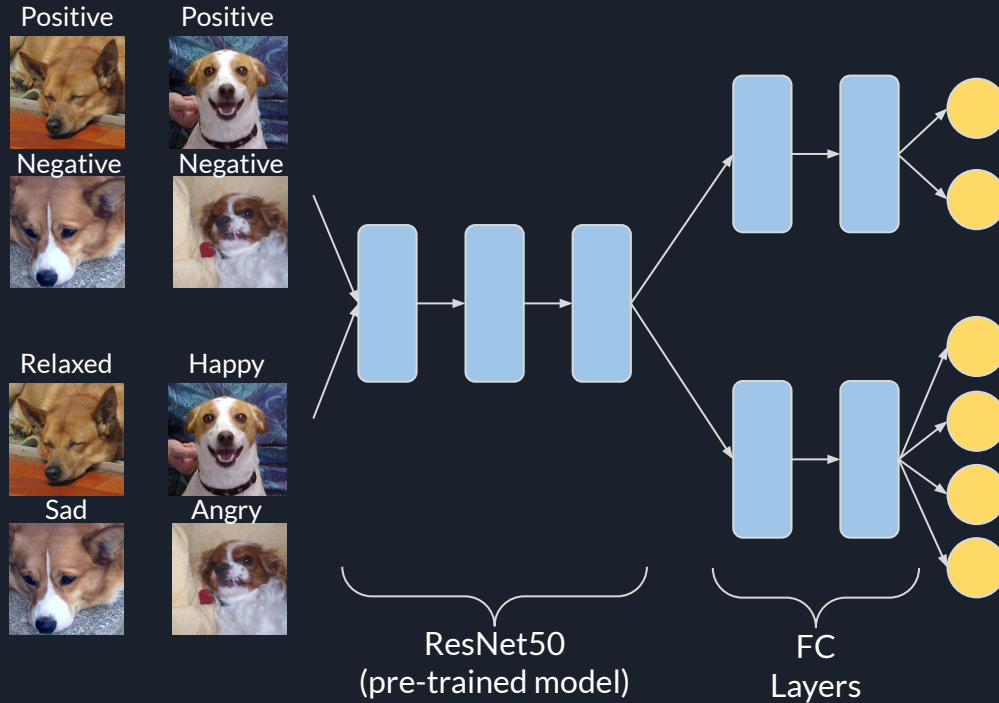
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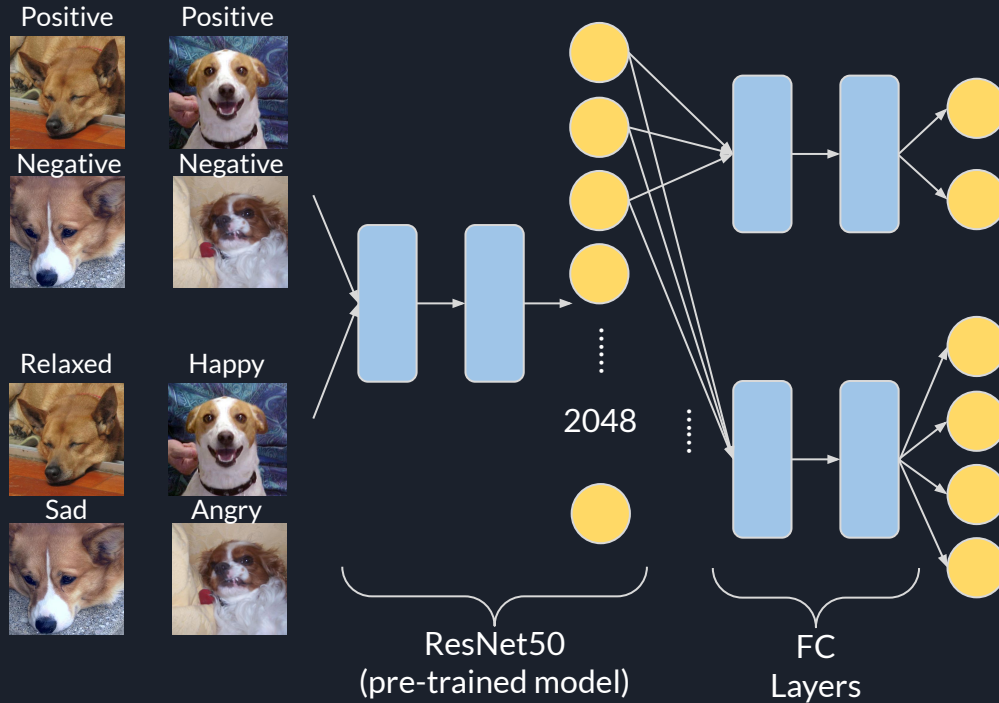
# Methodology (Multi Task Learning)



# Methodology (Multi Task Learning)



# Methodology (Multi Task Learning)





## Result (Multi Task Learning)

- 35%



# Methodology

- Classifier Modification
- Multi Task Learning
- **Fine-tuning**
- Unsupervised Domain Adaptation

# Methodology (Fine-Tuning)

**Big Idea:** Low-Level features are general and High-level features are more specific.

- Low-Level features learnt are **transferable**
- Higher Layers capture features specialized to the original training tasks

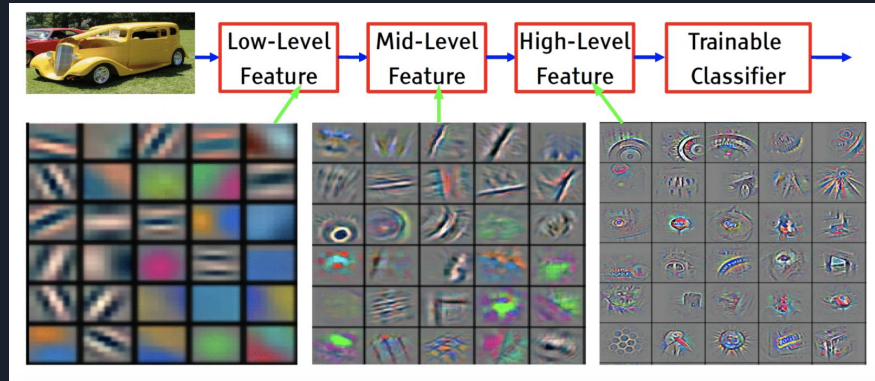


Figure 1: [Hierarchy of Features in CNNs](#)



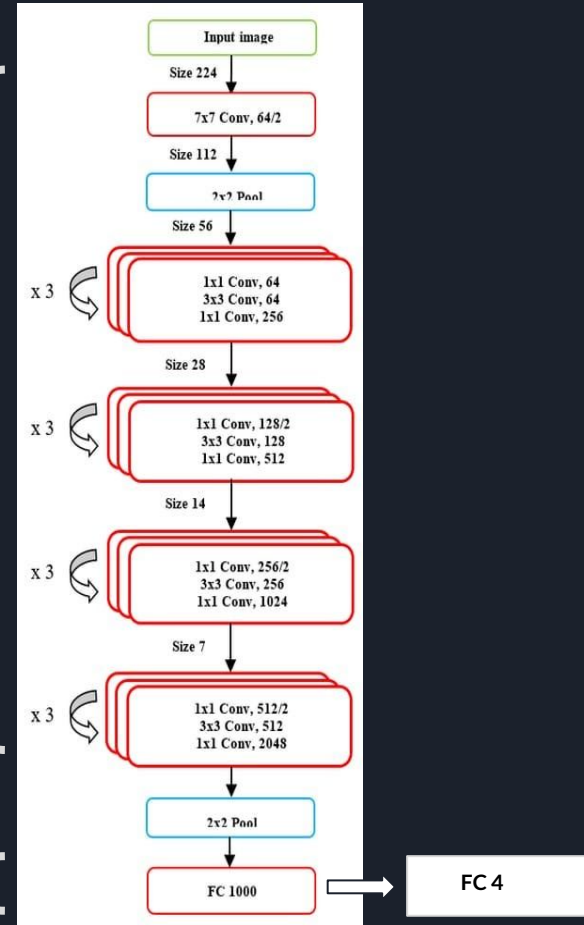
# Methodology (Fine-Tuning)

## ResNet50

- Previously: freeze all feature layers
  - Train only classification layer

Freeze

Train



# Methodology (Fine-Tuning)

## ResNet50

- Previously: freeze all feature layers
  - Train only classification layer
- Higher layer may capture features that are specialized for ImageNet task
  - Train last Feature Layer & FC Layer
- Use initial pre-train weights leads to better performance and generalization (Yosinski)

Jason Yosinski, Jeff Clune, Yoshua Bengio, & Hod Lipson. (2014). How transferable are features in deep neural networks?

Freeze

Train

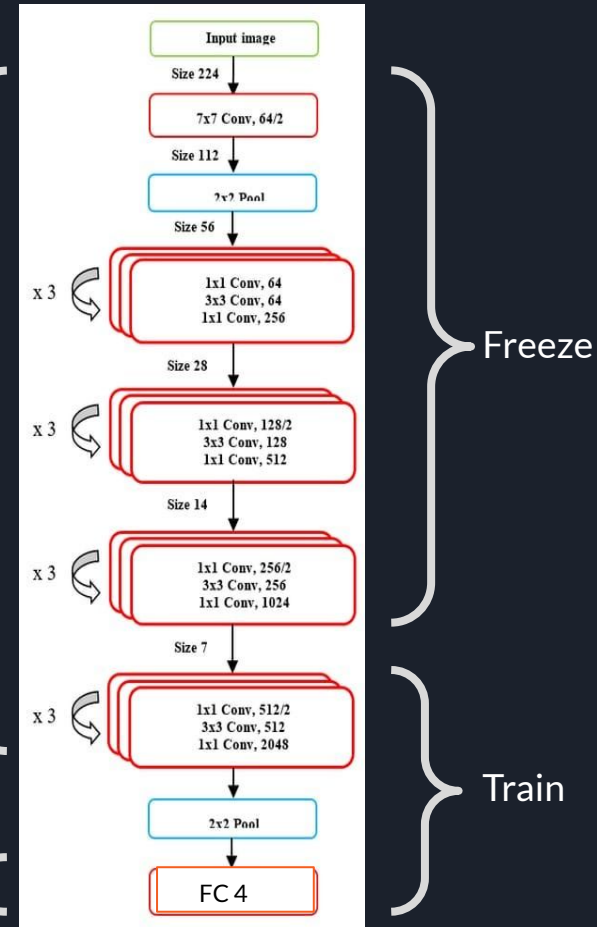


Image Credit: [ResNet50 Architecture](#)



## Results (Fine-Tuning)

Method	Training on FC Layer	Training on FC + Stage 4	Accuracy Increase
Training Accuracy	54.5%	63.8%	9.3%
Test Accuracy	53.3%	57.6%	4.3%
Accuracy Difference	-1.2%	-6.2%	



# Methodology

- Classifier Modification
- Multi Task Learning
- Fine-tuning
- Unsupervised Domain Adaptation

# Methodology (Unsupervised Domain Adaptation)

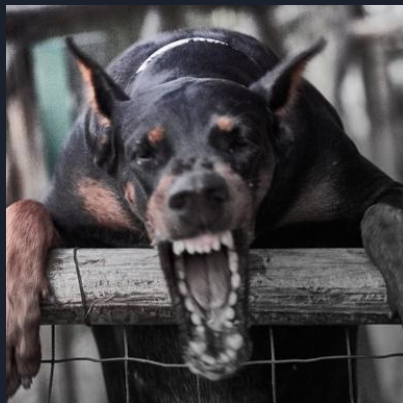


## What is Domain Adaptation

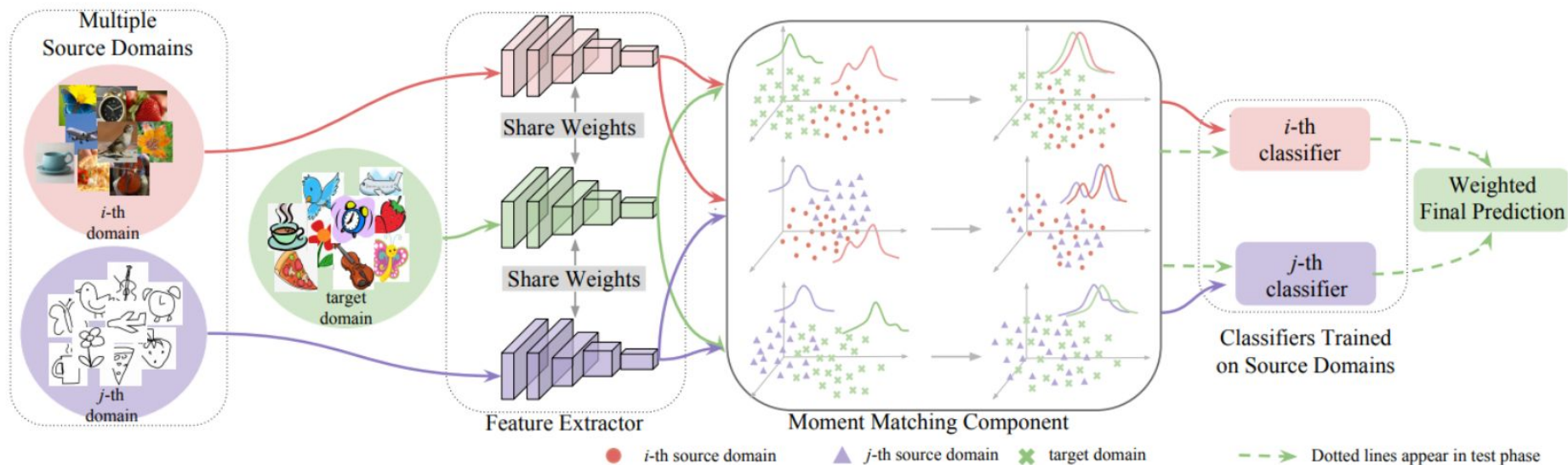
- Source domain  $\rightarrow$  target domain
- Two features: brightness, raindrop
- Bright + No raindrop = Sunny
- Dark + Rain drop = Rainy
- Dark + No raindrop = ?
- simple/single knowledge  $\rightarrow$  complex situation

## Why Domain Adaptation useful

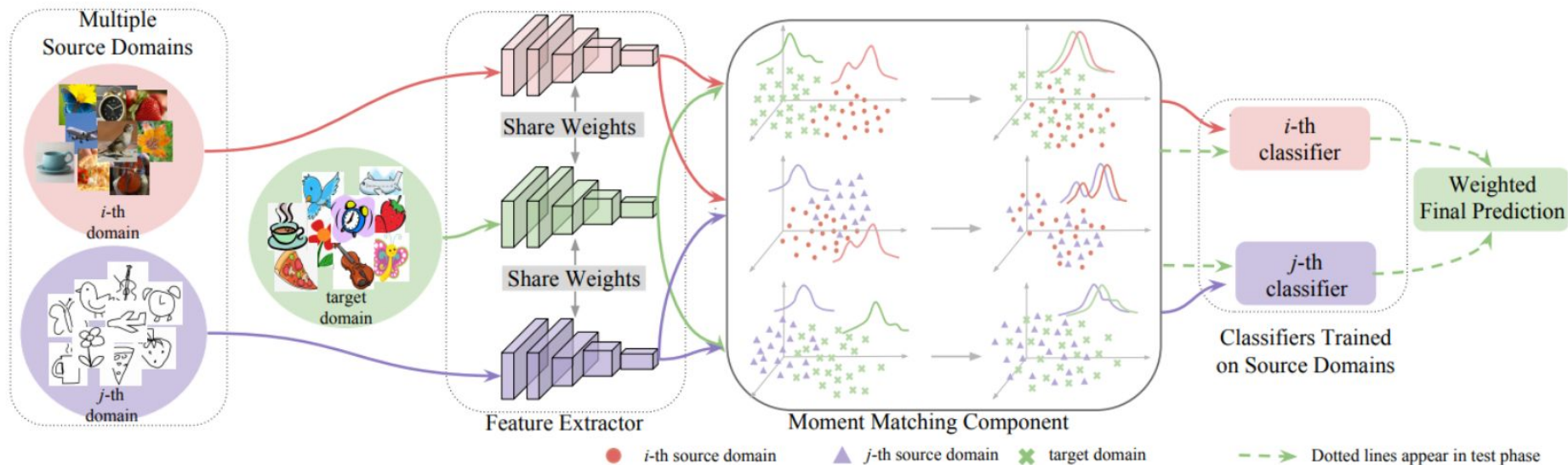
- Pre-trained model is trained on a specific dataset
- Our dataset contains real world pictures



# Methodology (Unsupervised Domain Adaptation)



# Methodology (Unsupervised Domain Adaptation)





# Results(Unsupervised Domain Adaptation)

## ResNet50 Accuracy

Training Accuracy: 62.8%

Testing Accuracy: 53%

Training Accuracy Increase: 9.8%





# Conclusion

- Introduced Pre-train Models and Transfer Learning
- Selected best Pre-train Model
- Classifier Modification
- Multi-task Learning
- Fine-Tuning
- Unsupervised Domain Adaptation