



Tennis Image Classification

Rebecca Ruppel
Galvanize Data Science Immersive Capstone

Supervised Learning

Labeled
Data

Train Model

Predictions on
Unlabeled Data

Random Forests

Neural Networks

Support Vector Machines

Naive Bayes

Supervised Learning

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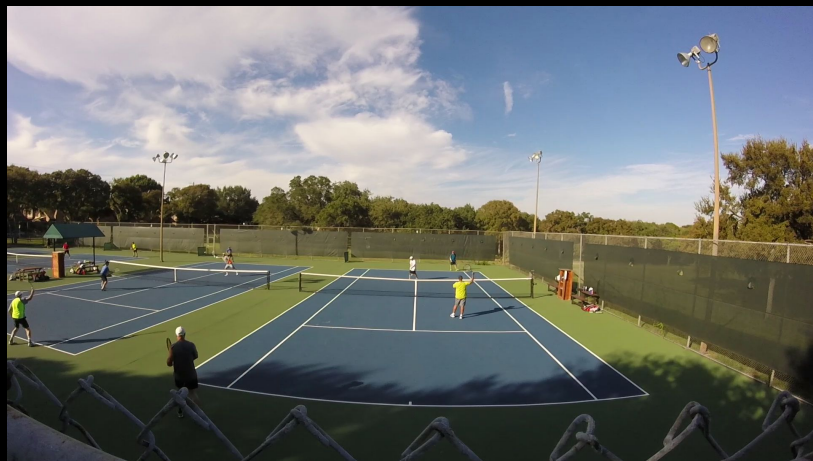
Random Forests

Neural Networks

Support Vector Machines

Naive Bayes

Cizr Data



Outside Point

Cizr Data



Outside Point



Serve

Cizr Data



Outside Point

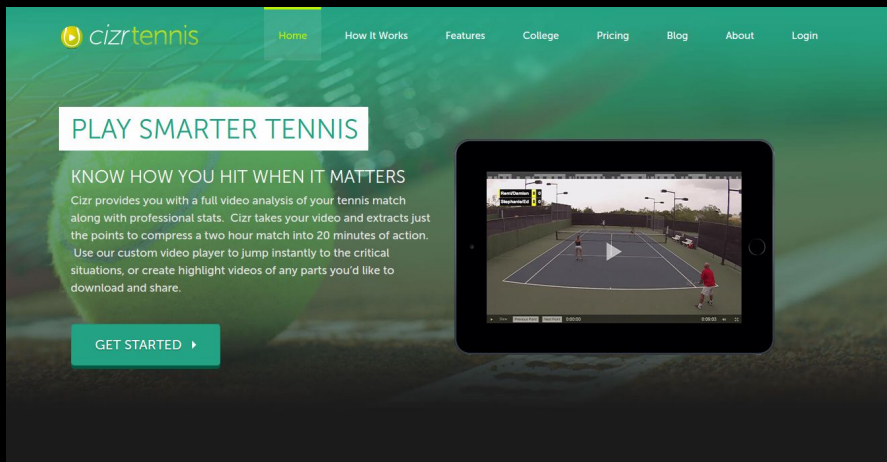


Serve



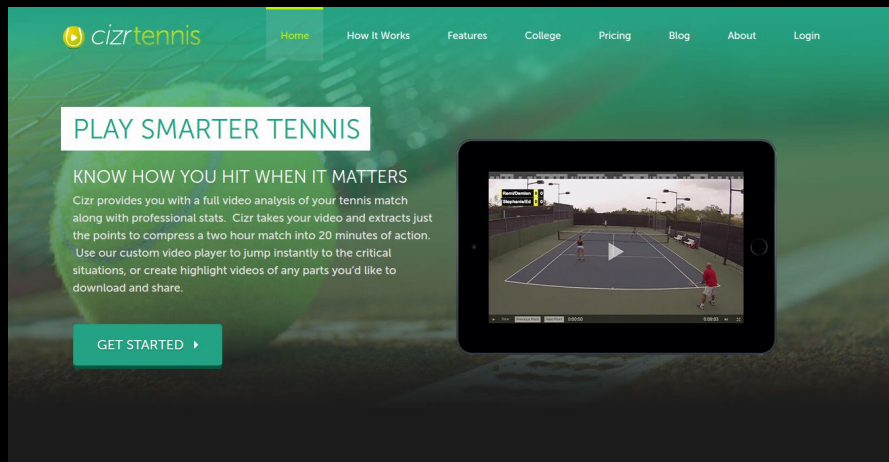
Inside Point

Cizr: Tennis video classification



- Tags tennis video to separate active play from rest of video
- Processes user video into active play
- Video data is categorized by hand

Cizr: Tennis video classification

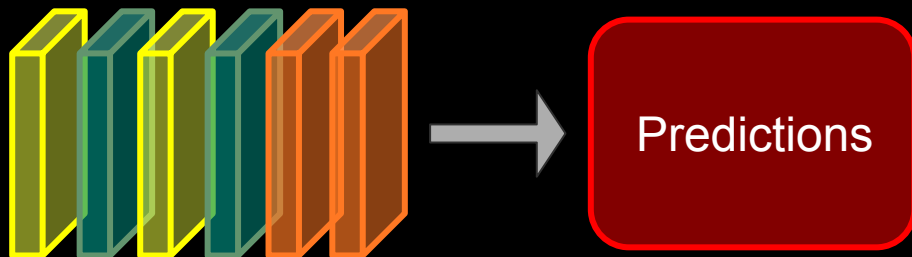


- Tags tennis video to separate active play from rest of video
- Processes user video into active play
- Video data is categorized by hand

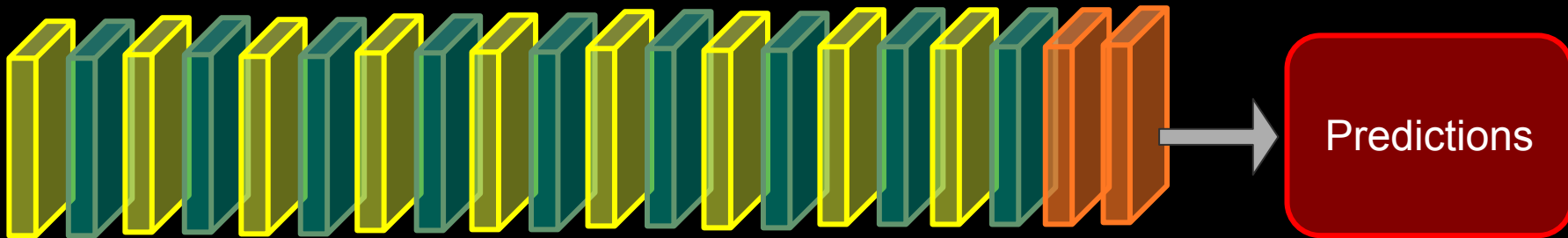
Neural network could be used to automate classification process

Deep Learning

Neural Network

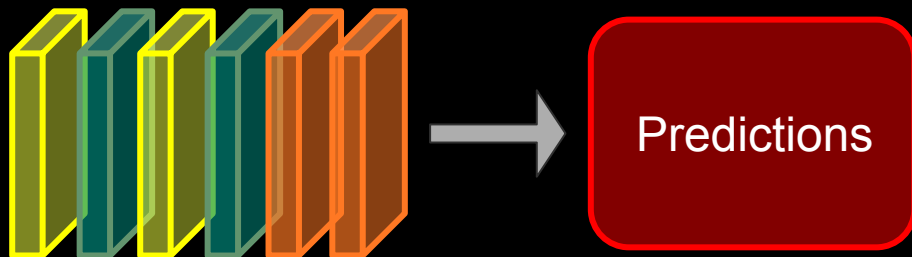


Deep Neural Network

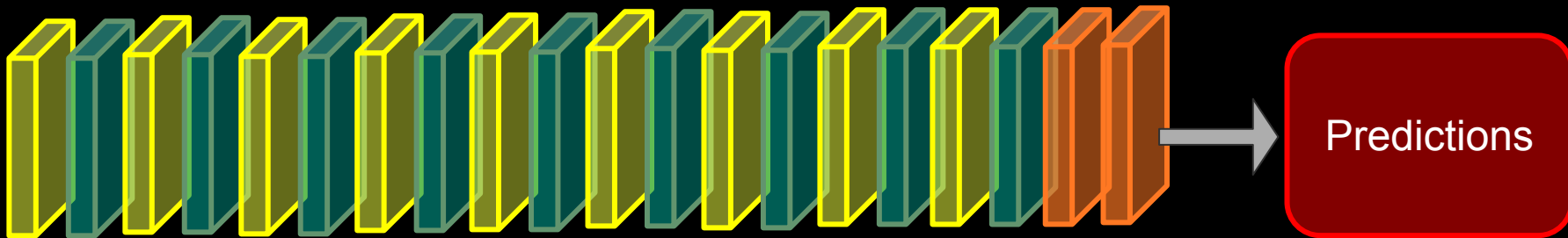


Deep Learning

Neural Network



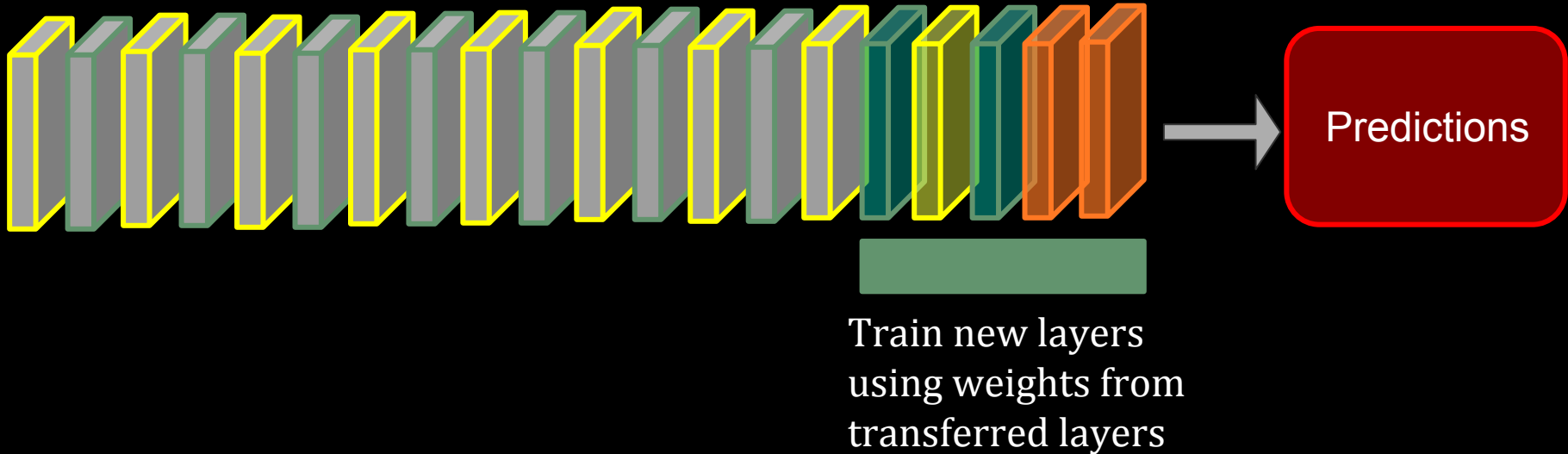
Deep Neural Network



Sadly, these take weeks to train on large datasets...

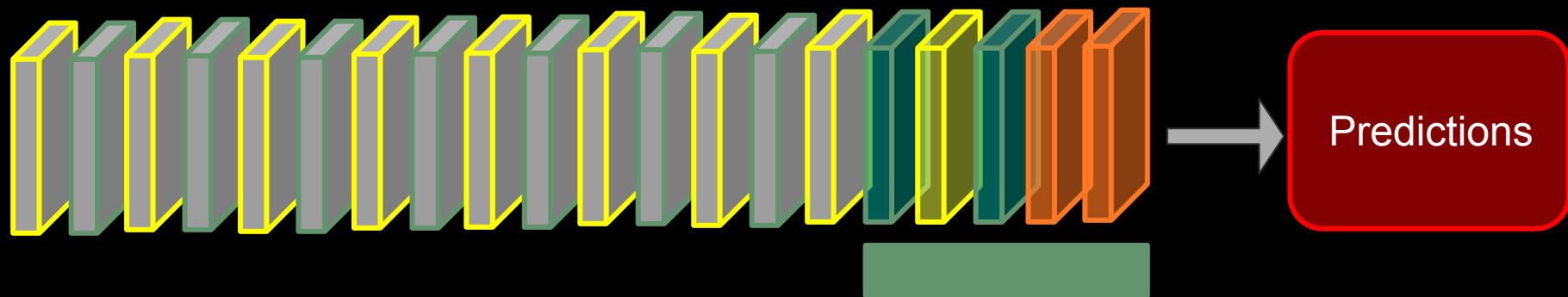
Deep Learning & Transfer Learning

Freeze Weights for Pretrained Layers



Deep Learning & Transfer Learning

Freeze Weights for Pretrained Layers



Train new layers using weights from transferred layers

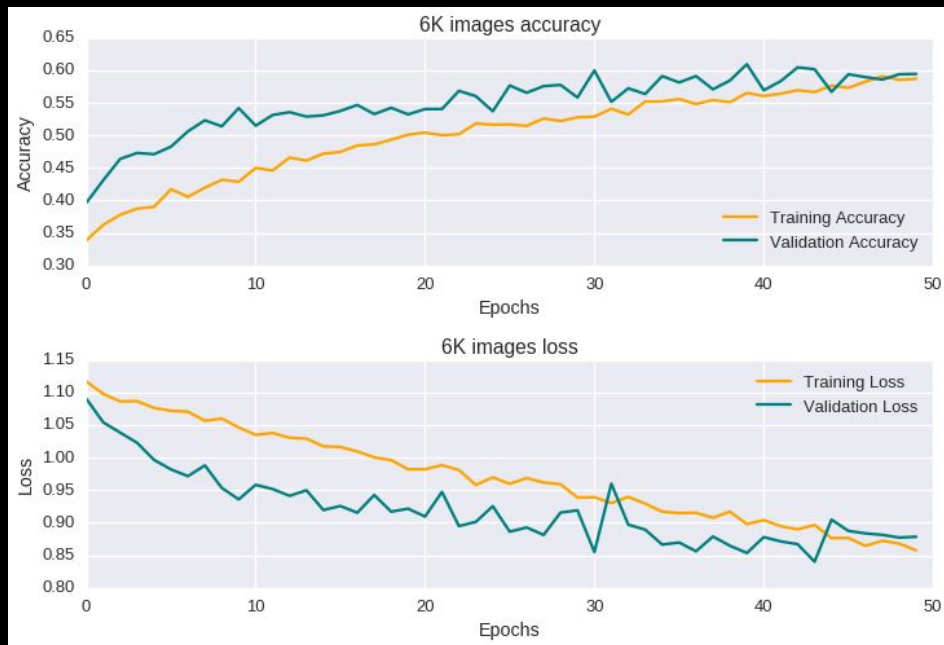
VGG16 ImageNet 2014 competition
Very Deep Convolutional Networks for Large-Scale Image Recognition
K. Simonyan, A. Zisserman
arXiv:1409.1556

Model Performance

6,000 images
4,800 Train/2,200 Validation

Labels

Predicted			
	Inside Point	Serve	Outside Point
Inside Point	347	222	122
Serve	82	564	59
Outside Point	139	204	338



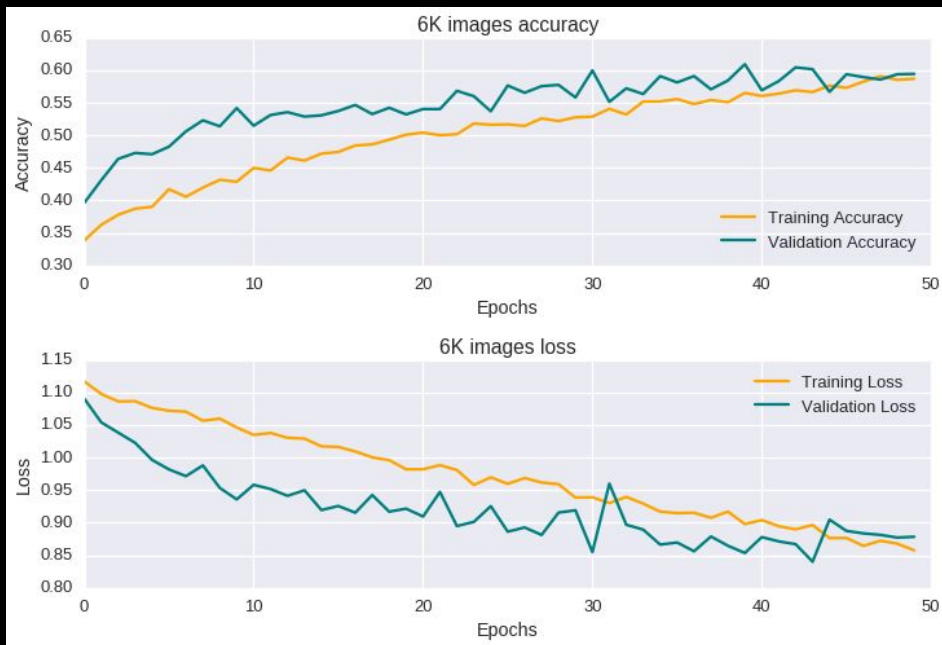
Validation Accuracy: 60%

Model Performance

6,000 images
4,800 Train/2,200 Validation

Predicted

Labels	Predicted		
	Inside Point	Serve	Outside Point
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Validation Accuracy: 60%

Model Performance

21,000 images
14,700 Train/6,300 Validation

Labels

Predicted			
	Inside Point	Serve	Outside Point
Inside Point	1333	344	486
Serve	272	1498	340
Outside Point	310	350	1367



Validation Accuracy: 66%

Model Performance

21,000 images
14,700 Train/6,300 Validation

Labels

		Predicted		
		Inside Point	Serve	Outside Point
Labels	Inside Point	1333	344	486
	Serve	272	1498	340
	Outside Point	310	350	1367



Validation Accuracy: 66%

Using Video in Neural Networks



Outside Point



Inside Point

Using Video in Neural Networks



Outside Point



Inside Point

Questions?

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