Form Check

Squat Classification with 3D Convolutional Neural Network

James Nikiforuk

"There is simply no other exercise, and certainly no machine, that produces the level of central nervous system activity, improved balance and coordination, skeletal loading and bone density enhancement, muscular stimulation and growth, connective tissue stress and strength, psychological demand and toughness, and overall systemic conditioning than the correctly performed full squat."

- Mark Rippetoe

A Solid Foundation

Squats are a foundational movement pattern. While proper execution has many benefits, improper execution can lead to strain, disability, and loss of work.

What's the cost?









CLAIMS MADE:

DAYS LOST:

DOLLAR COST:

%TOTAL:

151,752 Claims 2016-2020

8,319,279 Days 2016-2020

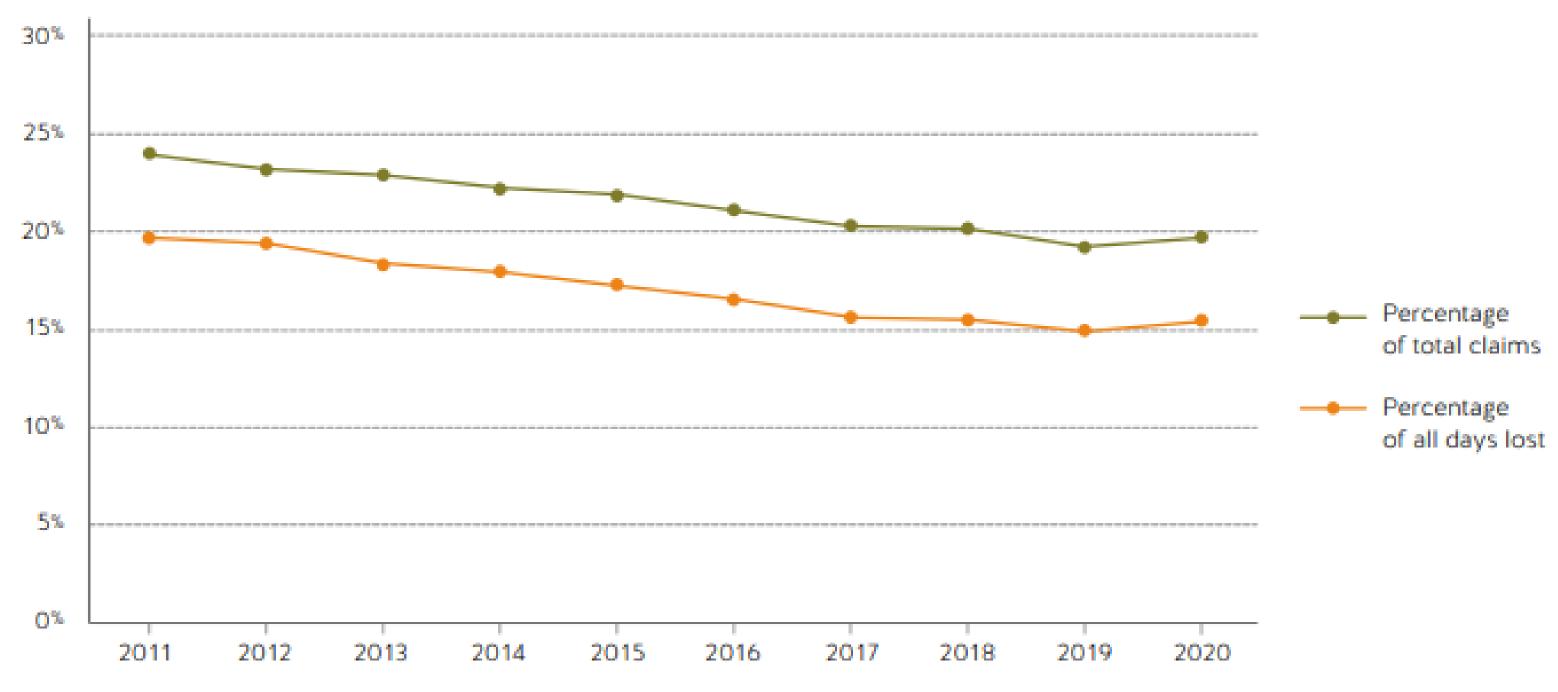
\$1,869,042,089 2016-2020

55.8% of Claims 2016-2020

Data from WorkSafeBC Statistics 2020 Report 'Strain, Back', 'Strain, Other', 'Hernia'

Back Strain

Back Strain claims as % of total claims and days lost.



Dataset and Approach

2000+ VIDEOS

Good squats and 6 common errors with varied angles and settings

SPLIT VIDEOS INTO FRAMES

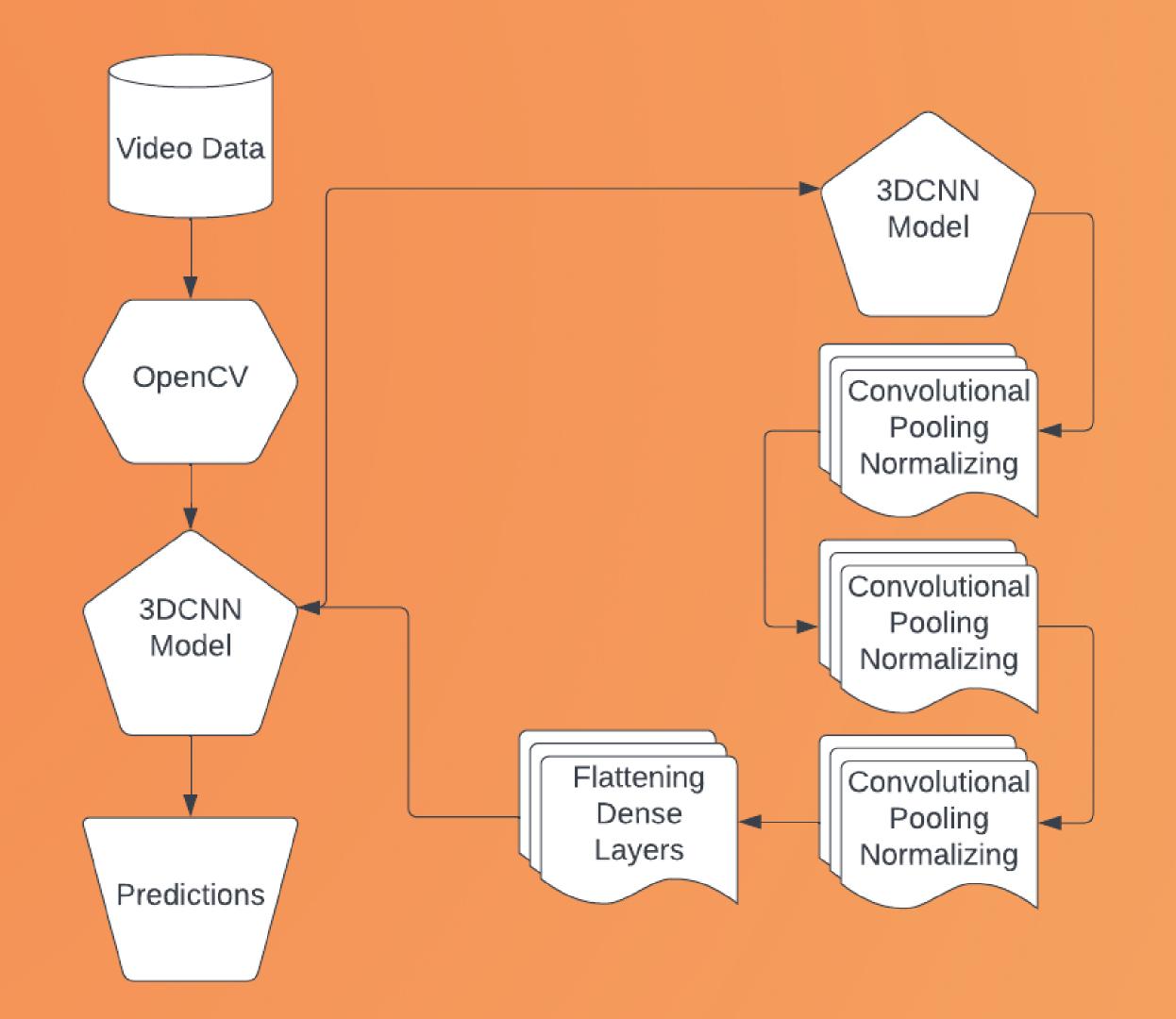
Create images for use with model

TRAIN MODEL ON IMAGES

Feed image data to 3DCNN architecture.

TEST MODEL WITH VIDEO

Evaluate trained model on unseen video



Performance Evaluation

BASELINE

14.65%

BEST ACCURACY

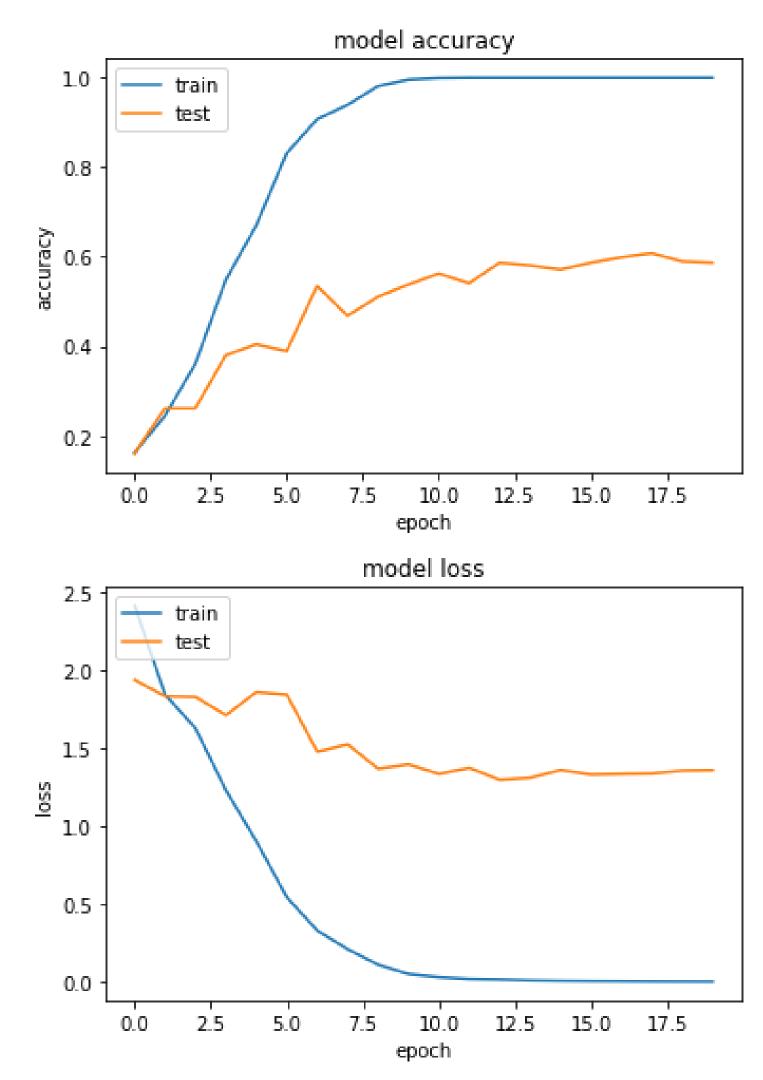
60.79%

BEST PRECISION BY CLASS

81.07%

Accuracy and Loss

Training and Validation metrics across Epochs



Class Confusion



WARPED BACK

Without Pose data, hard to differentiate Z-axis



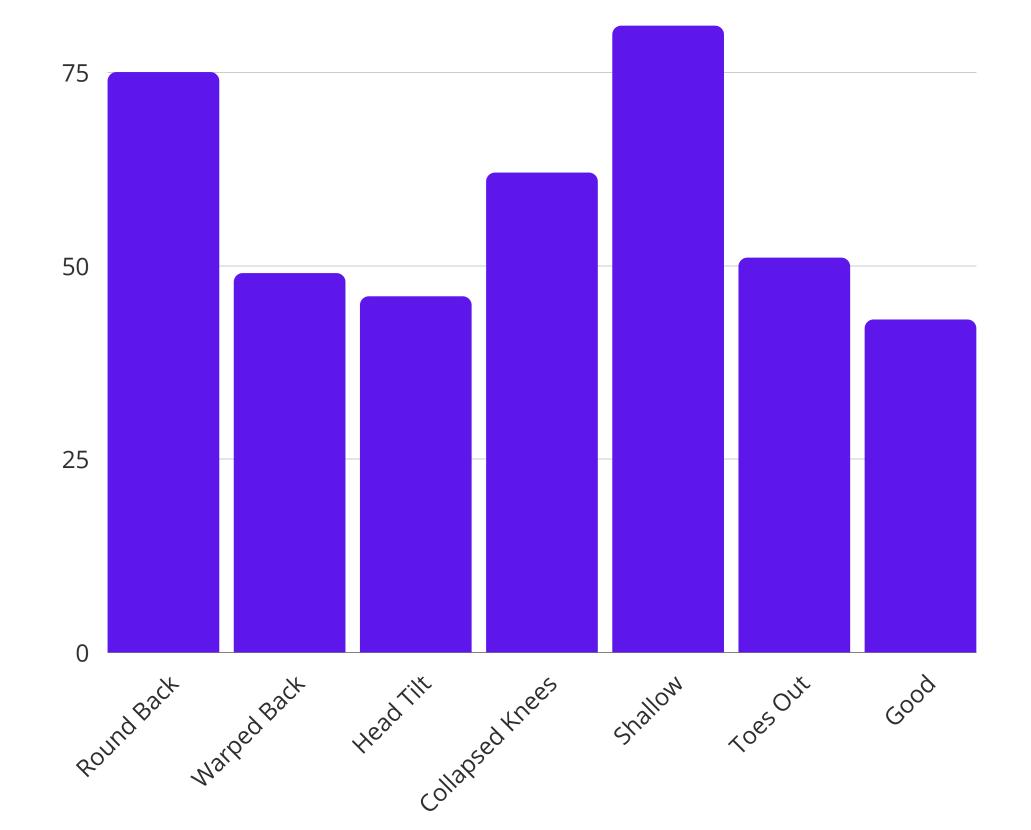
GOOD

Very similar to Warped Back from same angle

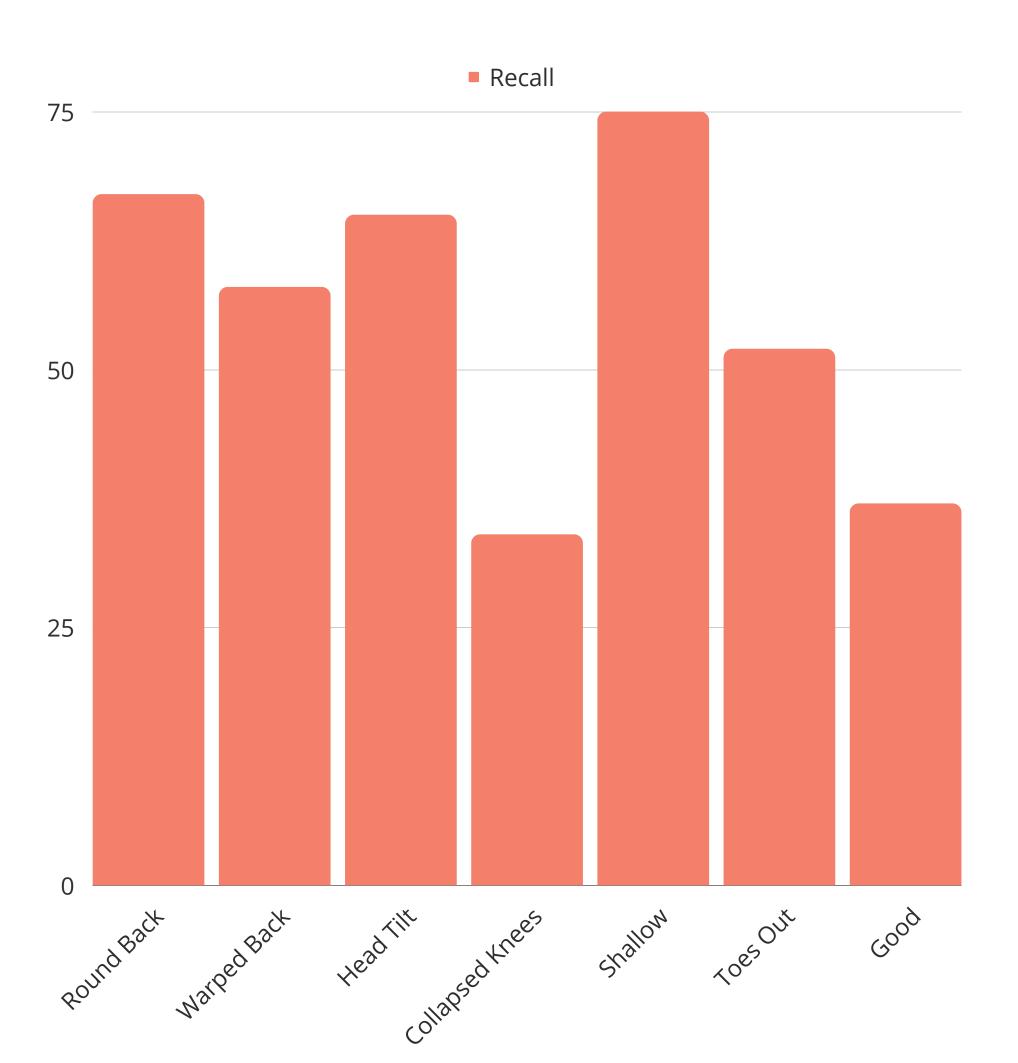


SHALLOW

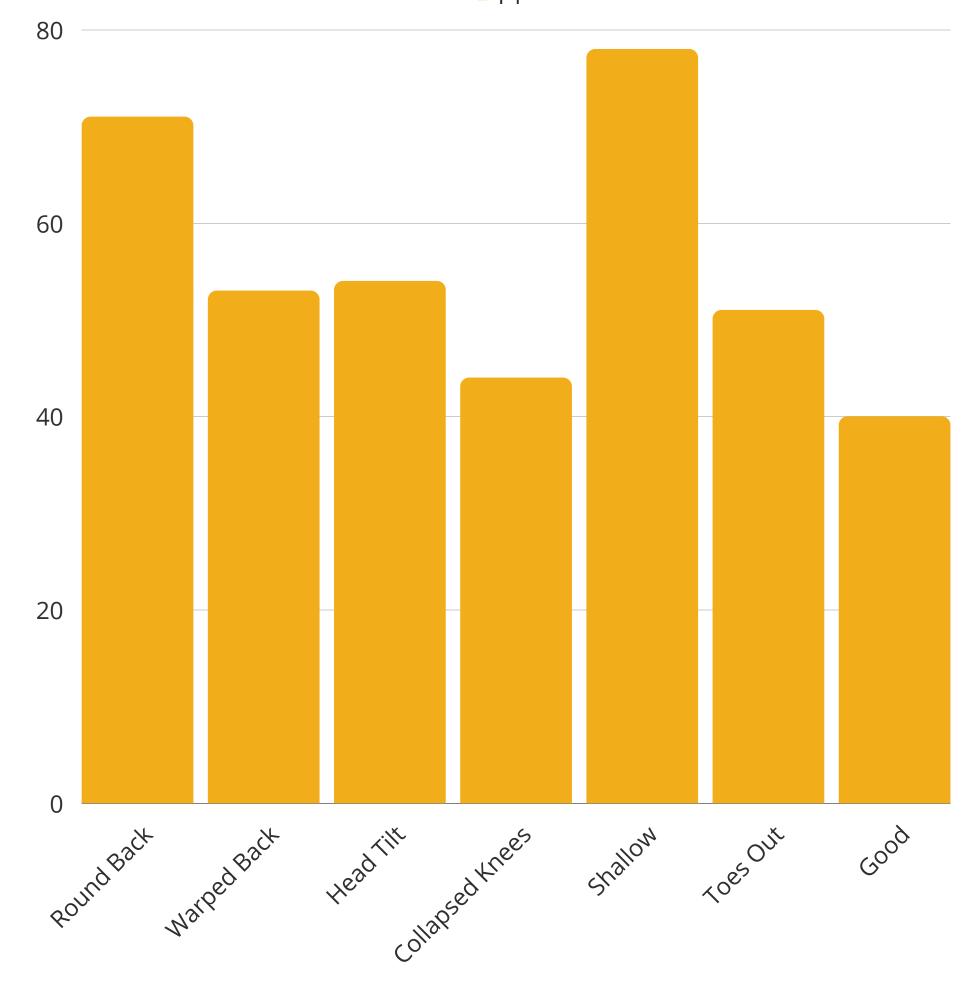
A much more distinct class



Precision by Class



Recall by Class



F1 Score by Class

Future Improvements

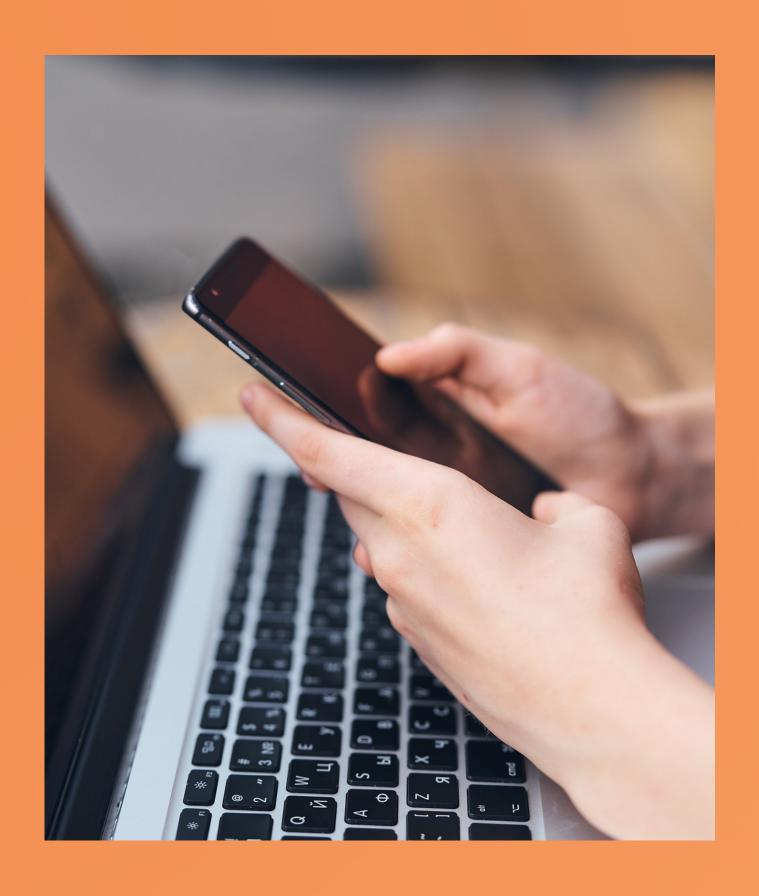
Techniques to implement for increased performance



PRE-TRAIN
MODEL ON
IMAGE DATA

UTILIZE POSE ESTIMATION DATA

FURTHER
ADJUSTMENTS TO
ARCHITECTURE



Thank you for listening!

Please reach out for any additional questions or comments

Appendix

APPENDIX A

Links to video and WorkSafe data

APPENDIX B

Model summary for 3DCNN architecture

APPENDIX C

Overall Class balance

APPENDIX D

Classification Report

VIDEO DATASET

http://hi.cs.waseda.ac.jp/~ogata/Dataset.html

Appendix A

WORKSAFE BC STATISTICS

https://www.worksafebc.com/en/resources/about-us/annual-report-statistics/2020-stats?lang=en

Appendix B

Layer (type)	Output Shape	Param #
conv3d_40 (Conv3D)		5248
<pre>max_pooling3d_30 (MaxPoolin g3D)</pre>	(None, 29, 31, 31, 64)	0
<pre>batch_normalization_30 (Bat chNormalization)</pre>	(None, 29, 31, 31, 64)	256
conv3d_41 (Conv3D)	(None, 27, 29, 29, 128)	221312
max_pooling3d_31 (MaxPoolin g3D)	(None, 13, 14, 14, 128)	0
batch_normalization_31 (Bat chNormalization)	(None, 13, 14, 14, 128)	512
conv3d_42 (Conv3D)	(None, 11, 12, 12, 256)	884992
conv3d_43 (Conv3D)	(None, 11, 12, 12, 512)	131584
<pre>max_pooling3d_32 (MaxPoolin g3D)</pre>	(None, 11, 12, 12, 512)	0
<pre>batch_normalization_32 (Bat chNormalization)</pre>	(None, 11, 12, 12, 512)	2048
dense_34 (Dense)	(None, 11, 12, 12, 256)	131328
flatten_16 (Flatten)	(None, 405504)	0
dense_35 (Dense)	(None, 512)	207618560
dense_36 (Dense)	(None, 7)	3591

Total params: 208,999,431 Trainable params: 208,998,023 Non-trainable params: 1,408

Appendix C

	Videos	%
good	311	15.07
bad_back_round	288	13.95
bad_back_warp	320	15.50
bad_inner_thigh	235	11.39
bad_toe	305	14.78
bad_shallow	325	15.75
bad_head	280	13.57

Appendix D

	precision	recall	f1-score	support
bad_back_round	0.75	0.67	0.71	49
bad_back_warp	0.49	0.58	0.53	69
bad_head	0.46	0.65	0.54	57
bad_inner_thigh	0.62	0.34	0.44	44
bad_shallow	0.81	0.75	0.78	72
bad_toe	0.51	0.52	0.51	58
good	0.43	0.37	0.40	62
accuracy			0.56	411
macro avg	0.58	0.55	0.56	411
weighted avg	0.58	0.56	0.56	411