Deep Learning for Emotion Recognition in Cartoons

Under the guidance of

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Aim

MEASURE HOW ACCURATE THE PROGRAM IS ABLE TO IDENTIFY AN EMOTION FROM A GIVEN CARTOON VIDEO.

REQUIREMENTS

- Choose a cartoon: Choice was Tom & Jerry².
 - Lots of various emotions
 in each episode.
- Segment faces from the cartoon.
- Build a dataset of emotions for each main character (Tom & Jerry)
- Train the network on a labeled dataset.



² Tom & Jerry © Warner Bros. Entertainment, Inc

DATASET GATHERING

Tom & Jerry Q

About 4,260,000 results = FILTER



Tom and Jerry, 88 Episode - Pet Peeve (1954)

Jonni Valentayn @ 1.4M views • 1 week ago

The short three-minute fragment from series is the 88th one-reel animated Tom and Jerry short, directed by William Hanna and ...



Tom and Jerry, 89 Episode - Touché, Pussy Cat! (1954)

Jonni Valentayn @ 387K views • 5 days ago

The short three-minute fragment from series is a 1954 one-reel animated Tom and Jerry short, directed by William Hanna and ...



Tom and Jerry, 87 Episode - Downhearted Duckling (1954)

Jonni Valentayn @ 1.6M views • 3 weeks ago

The short three-minute fragment from series is the 87th one reel animated Tom and Jerry cartoon released. It was created in 1953, ...



Jonni Valentayn @

829,643 subscribers · 99 videos

Tom and Jerry - is an American animated series of short films crea...

SUBSCRIBE 829K



Tom and Jerry, 88 Episode -Pet Peeve (1954)

1.4M views • 1 week ago

Tom and Jerry, 13 Episode -

168K views • 1 month ago

The Zoot Cat (1944)



Tom and Jerry, 87 Episode -Downhearted Duckling

1.6M views • 3 weeks ago

Tom and Jerry, 11 Episode -

The Yankee Doodle Mouse

296K views • 1 month ago



Tom and Jerry, 86 Episode -Neapolitan Mouse (1954)

2.1M views • 1 month ago



Tom and Jerry, 85 Episode -Mice Follies (1954)

1M views • 1 month ago



Tom and Jerry, 10 Episode -The Lonesome Mouse (1943)

270K views • 1 month ago



Tom and Jerry, 9 Episode -Sufferin' Cats! (1943)

190K views • 1 month ago



Tom and Jerry, 3 Episode -



Tom and Jerry, 6 Episode -Tom and Jerry, 5 Episode -Dog Trouble (1942) Puss n' Toots (1942)



Tom and Jerry, 4 Episode -Fraidy Cat (1942)



The Night Before Christmas

Tom ar The Mi

Tom ar

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2.1M vi

Tom ar

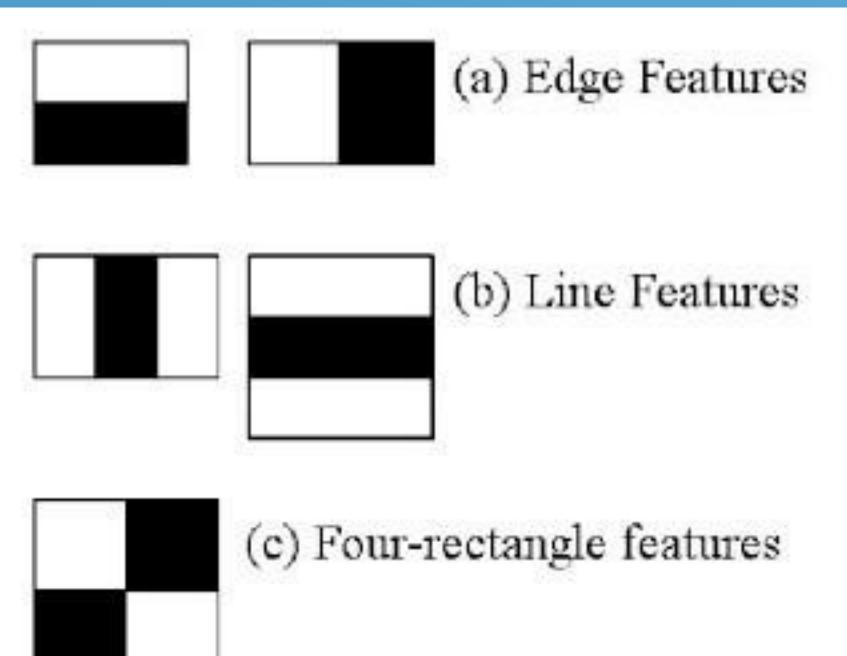
Fine Fe

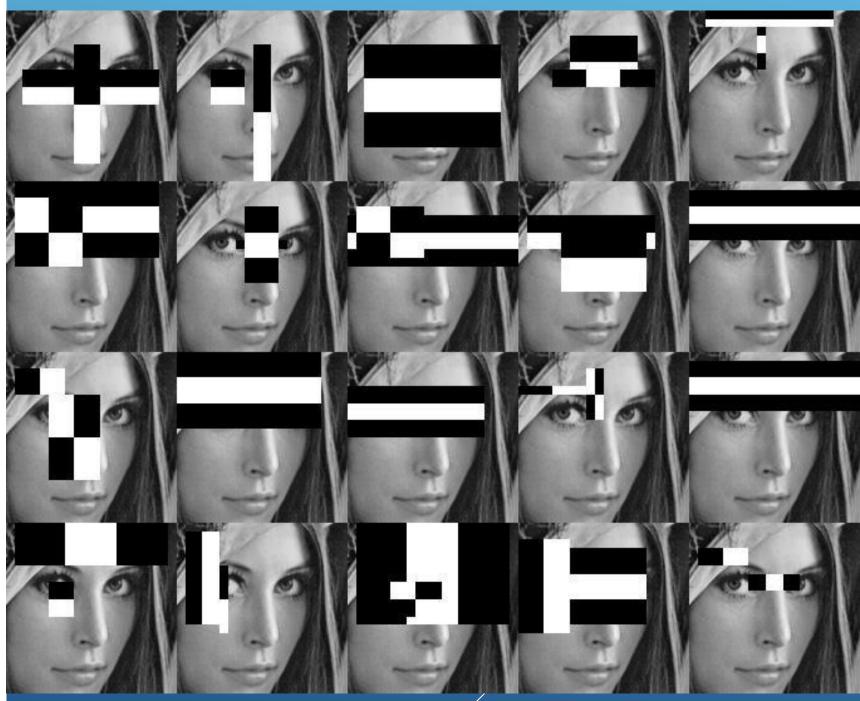
587K vi

SEGMENTATION

HAAR CASCADES

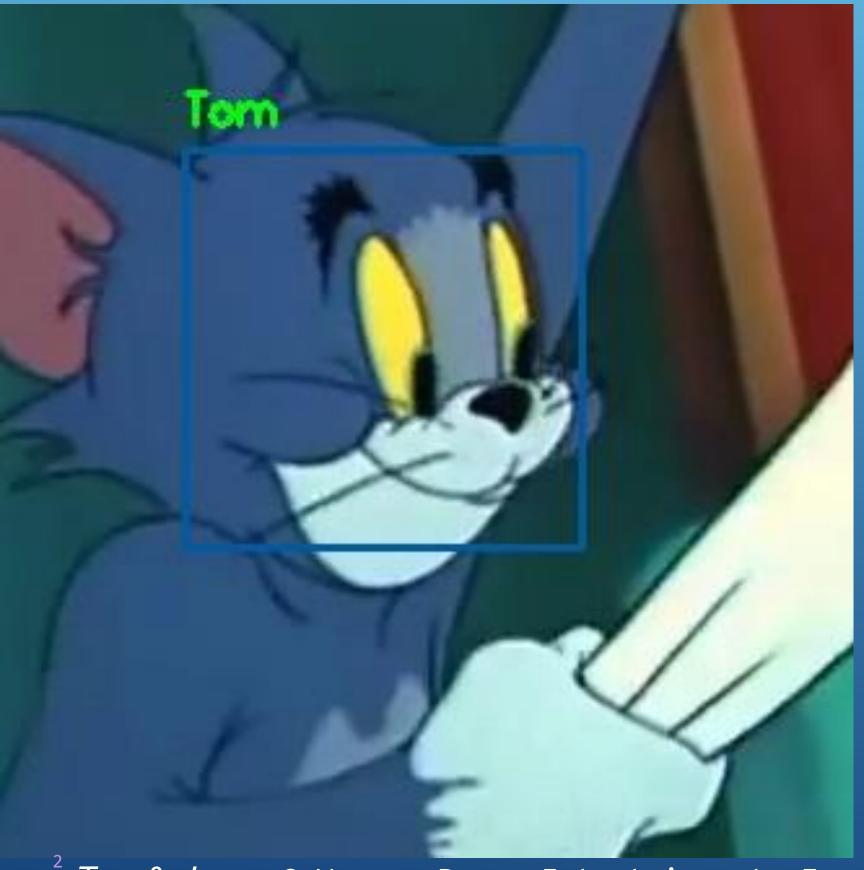
HAAR CASCADES



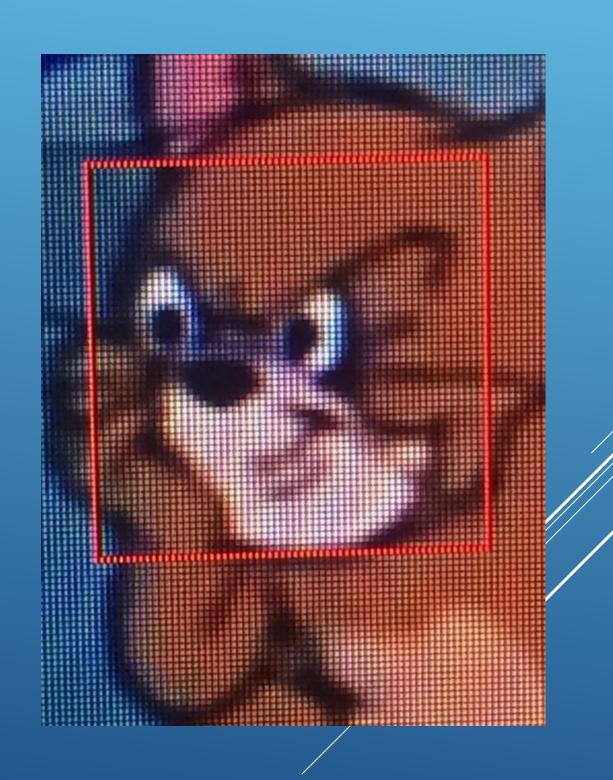


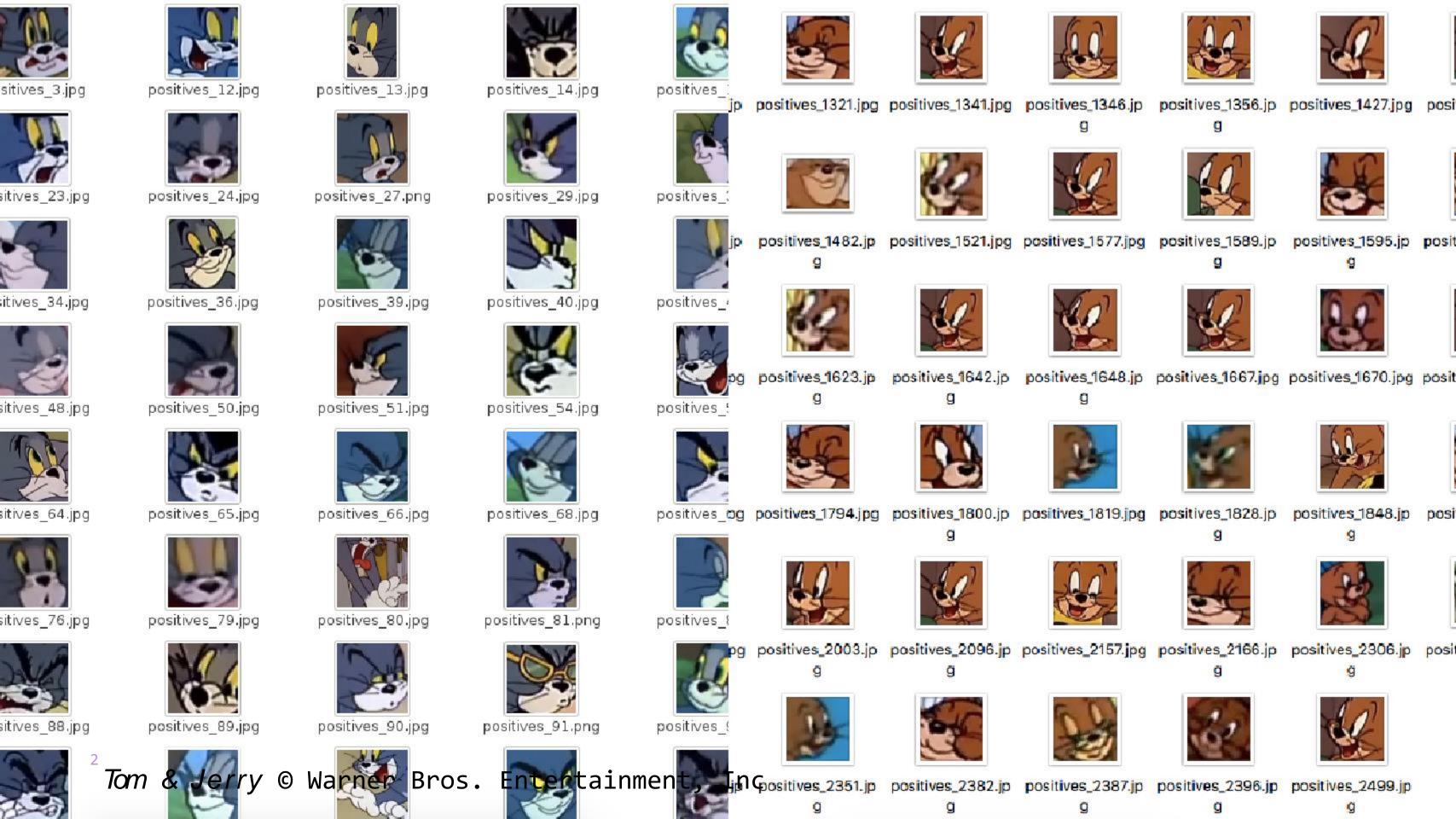
Haar Cascades

- Created custom Haar cascade for both Tom & Jerry.
 - There were none made online to detect cartoon faces, only human ones.
 - Depending on the window size it does detect other character faces in the cartoon.



Tom & Jerry © Warner Bros. Entertainment, Inc



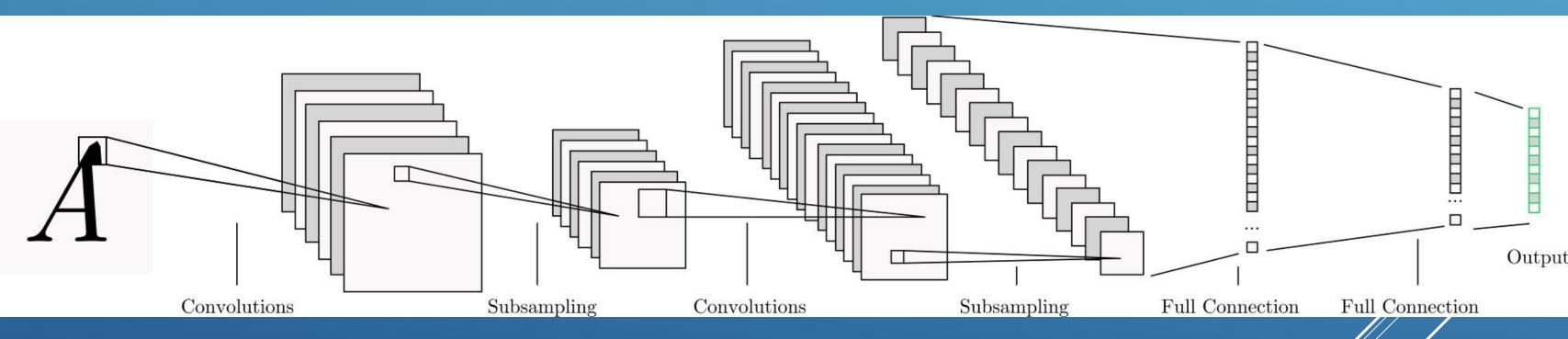


DATASET STATS

Dataset Stats

- In total about 159,035 images segmented.
 - For about ~64 episodes. (Tom & Jerry has over 100)
- Selected around 400 images for each character & emotion, (angry, happy, surprise) for training and testing.

CONVOLUTIONAL NERUAL NETWORK



Convolutional Nerual Network

- In recent years CNN's produced great results in image & object recognition.
- The CNN is used for this project to learn features (eg. smile angles, eyebrows).
- Framework for DL used was Keras + TensorFlow backend. (Keras also works with Theano)

Convolutional Nerual Network

- No pre-trained network.
 - Inception-V3 predicts Tom & Jerry as 'comic books'.
- Images resized to 60x60 with 3 channels. (RGB)
- 3x3 convolution & 2x2 max pooling with a input
 image of size 60x60x3.

Convolutional Nerual Network

- 3x3 Convolution & 2x2 Max Pooling. (ReLU activation)
- 3x3 Convolution & 9x9 Max Pooling. (ReLU activation)
- Fully connected layer of 512 neruons.
- Final output layer of 6 neruons for each emotion.

ALGORITHM USED

Algorithm Used

The optimization algorithms that are tested and evaluated in the artefact are the following:

- Stochastic Gradient Descent (SGD)
- Adagrad
- Adadelta
- RMSprop
- Adam

RESULTS

Results

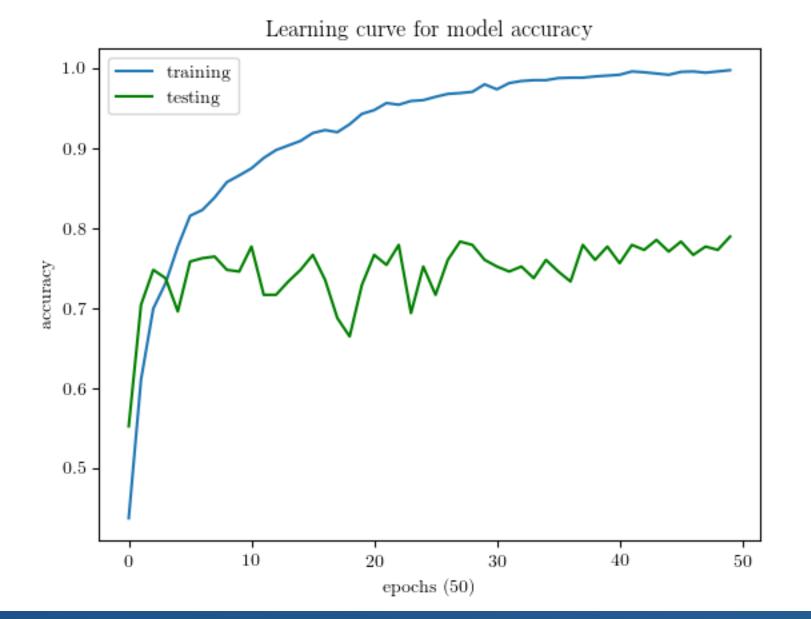
- Split dataset into 80% training, 20% testing.
- Trained network for 50 epochs on one GPU (Nvidia).
- Tested 5 optimisers for 5 runs:
 - Adadelta, Adagrad, Adam, RMSprop & Stochastic
 Gradient Descent (SGD)
- Hyperparameters (Layer size, Max pooling size...)

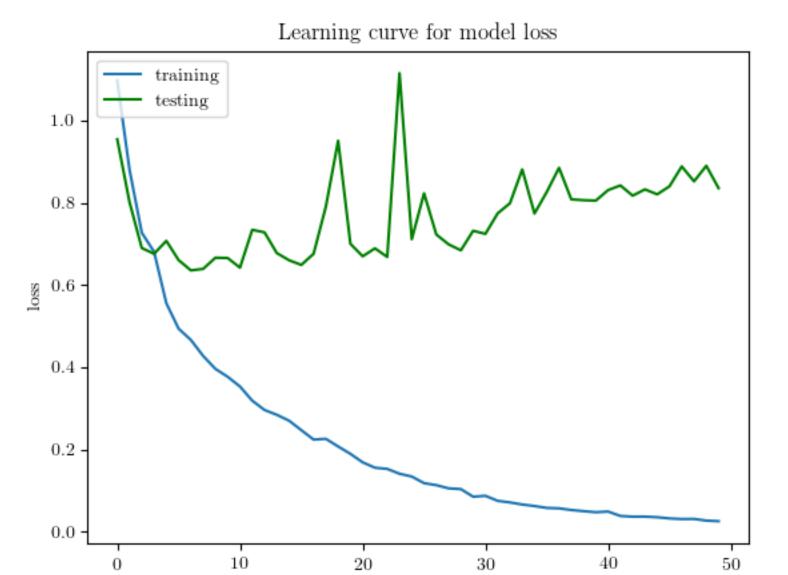
Results

- The network removes around 20-50% of neurons from the network when training. (Dropout)
 - Prevents overfitting the network.
- Rmsprop overfits the network.
- Adadgrad tends to underfit the network slightly.

Results

- Adadelta & SGD optimiser works well with slight overfitting.
- Adam has comparable performance to SGD but underfits in some test runs.
 - Adadelta was the best, but SGD was better for 3 test runs. (Both achieved ~80% accuracy)





epochs (50)

Parameter	Value	
1st Layer	3x3 Convolution	
2nd Layer	2x2 Maxpooling	
Dropout	20%	
Neurons in 1st FC Layer	512	
Neurons in 2nd FC Layer	6	
Metric	Categorical Cross-Entropy	
Epochs	50	

Table: Hyperparameter Table, Run 1.

Algorithm	Model Loss	Model Accuracy
Adadelta	1.3%	0.8%
Adagrad	10.7%	0.3%
Adam	1.3%	0.8%
RMSProp	10.7%	0.3%
SGD	0.8%	0.8%

Table : Results Table, Run 1

Run	Algorithm	Model Loss	Model Accuracy
1	SGD	0.8%	0.8%
2	SGD	0.9%	0.8%
3	SGD	1.0%	0.8%
4	Adadelta	1.0%	0.7%
5	Adadelta	0.6%	0.8%

Table : Algorithms with the best loss accuracy out of all 5 runs

~80% MODEL ACCURACY ~90% MODEL LOSS

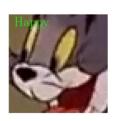
CLASSIFICATION RESULTS



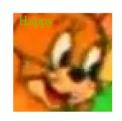
Happy 86.8% Surprise 9.5% Angry 3.6%



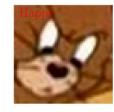
Happy 97.9% Surprise 1.5% Angry 0.6%



Happy 91.0% Surprise 5.5% Angry 3.5%



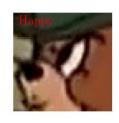
Happy 52.7% Angry 34.0% Surprise 13.3%



Happy 55.8% Surprise 33.2% Angry 10.8%



Happy 67.7% Angry 24.7% Surprise 7.8%



Happy 50.5% Angry 37.1% Surprise 14.7%



Angry 48.5% Happy 36.2% Surprise 15.3%



Angry 55.7% Happy 29.6% Surprise 14.7%





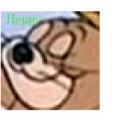
Surprise 99.4% Happy 0.5% Angry 0.1%



Angry 100.0% Happy 0.0% Surprise 0.0%



Surprise 99.9% Happy 0.1% Angry 0.0%



Happy 97.1% Angry 2.8% Surprise 0.1%



Happy 55.1% Angry 44.7% Surprise 0.2%



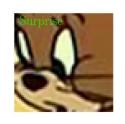
Angry 76.6% Happy 23.4% Surprise 0.0%



Surprise 99.3% Angry 0.5% Happy 0.2%



Happy 99.9% Surprise 0.1% Angry 0.0%



Surprise 88.3% Angry 8.8% Happy 2.9%

Stochastic Gradient Descent (SGD)

Potential Applications

- Animators
 - Automatic reference dataset.
 - Drawing -> Results of cartoons
 with similar emotions.
- Automatic subtitles.
- Recommendation Systems (Movies: which character is the happiest?)

References

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- Gajarla, V. and Gupta, A. (n.d) Emotion Detection
 and Sentiment Analysis of Images. In: cc.gatech.edu.

THANKYOU