

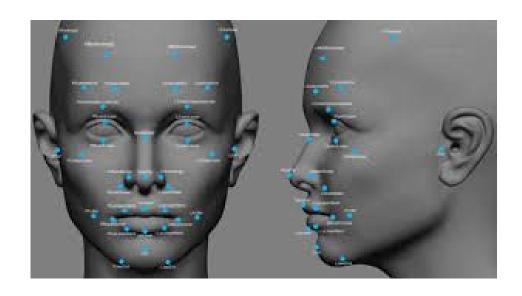
# Show Me Your Face!:)

**Metis Project 3: McNulty** 

Kevin Du Sarick Shah Nicholas Thomas Jennifer M. Wang

### Introduction

 Most humans can recognize others' emotions easily, but what about computers and machines?



#### Miscellaneous uses:

Security, privacy protection, digital personalization, entertainment, and even mental health benefits!



### **Goals**

- Build a model that will automatically recognize an emotion associated with a given image.
- Integrate model with a webcam application that displays a timeline of emotions.



### Method

### 1) FER2013 dataset - Facial Emotion Recognition (Kaggle)

Over 30,000 images, 48 x 48 pixels grayscale

### 2) Cohn-Kanade dataset (CK)

- ~ 700 images, 350 x 350 pixels grayscale
- Across both data sets, emotions include: neutral, angry, disgust, fear, happy, sad, surprise



### **Pipeline**

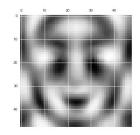
#### Face Recognition

 Template matching to detect features in faces (nose, mouth, eyes, etc.)



- Eigenfaces and Fisherfaces
- Find low dimension representation of a face which comes from a higher dimension
- A way to describe the most dominant features of the training set as a whole.







# **Results- Model Accuracy:**

	Cohn-Kanade (CK)	Kaggle FER2013
SVM	84%	44%
Fisher Face Recognizer	82%	32%
Multilayer Perceptron	77%	37%
RandomForest Classifier	54%	42%

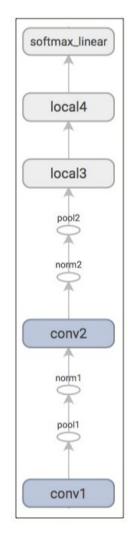
### **MLP Image Misclassification**



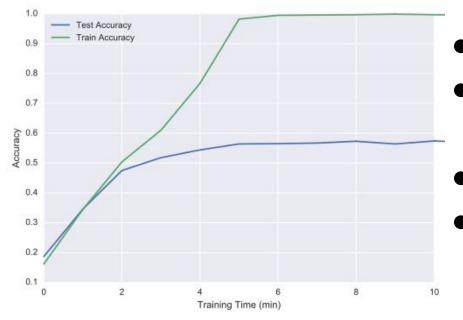
Classified as Anger--Actual label Neutral

### **Convolutional Neural Network**

- State of the art method for image classification
  - Automatically learns features
- Requires huge amount of data
  - Only trained on Kaggle FER2013 dataset
- Network from Tensorflow tutorial
  - Uses ReLU for activation functions



### **Convolutional Neural Network - Results**



Peaked after 6 minutes of GPU training 20x slower on CPU

- Training accuracy: ~ 0.999
- Test accuracy: ~ 0.57
  - Winning model: 0.711
- Extreme overfitting
- "Dropout" regularization by randomly dropping nodes from the neural net



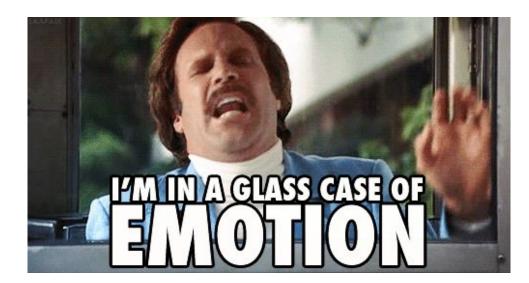
## **Conclusion and Implications**

- Ultimately, machines and computers are still not as good at detecting emotions compared with humans
  - Facial expressions are much more complex than the label of a single emotion.
  - Advances in machine facial recognition will yield tremendous societal benefits.



#### Potential uses for facial recognition data:

- Security, privacy, personalization of digital devices...etc.
- We are often unaware of how we *really* feel given how subtle our emotions may be. Facial recognition can help us better understand ourselves and attend to our needs.





### Machine facial recognition for improving mental health

- Mood monitoring (e.g., for depression)
- Efficacy of psychiatric drugs
- Advances in facial recognition may help improve compliance and the accuracy of these records





Emotions are universal and cross-cultural. Advances in machine facial recognition will help us better understand ourselves, each other, and our complex world.





### Resources

- <a href="http://eyalarubas.com/face-detection-and-recognition.html">http://eyalarubas.com/face-detection-and-recognition.html</a>
- http://docs.opencv.org/2.4/modules/contrib/doc/facerec/facerec\_tutorial.html
- <a href="http://www.paulvangent.com/2016/04/01/emotion-recognition-with-python-opency-and-a-face-dataset/">http://www.paulvangent.com/2016/04/01/emotion-recognition-with-python-opency-and-a-face-dataset/</a>
- https://realpython.com/blog/python/face-recognition-with-python/
- http://docs.opencv.org/trunk/d7/d8b/tutorial\_py\_face\_detection.html

# Fisherface/ Eigenface SUPPL.

#### Fisherfaces:

- Imagine a situation where the variance is generated from external sources, let it be light The axes with maximum variance do not necessarily contain any discriminative information at all, hence a classification becomes impossible.
  So a class-specific projection with a Linear Discriminant Analysis was applied to face recognition
- Minimize the variance within a class, while maximizing the variance between the classes at the same time.

#### Eigenfaces:

- Lower-dimensional subspace is found with Principal Component Analysis, which identifies the axes with maximum variance.
- Performance degrades with different lighting conditions