Project B Facial Emotion Recognition

ECE1512 Digital Image Processing and Applications
Department of Electrical and Computer Engineering
University of Toronto
Winter 2021

Karush Suri

Logistics

Project B uploaded

Deadline-Monday, April 5, 5:00 PM EST.

- Please start early!
- Questions..
 - FAQs page (primary source)
 - Discussion board
 - TA by email ("Project-B" as subject line)



- Tutorial (~ 1.5 hours)
 - Part-1: Facial Expression Recognition
 - Part-2: Facial Action Units
 - Part-3: Overview of PyTorch
- Break (~ 5 minutes)
- Open question hour (~ 0.5 hour)
 - Questions, questions, questions...

In case you have a question-

- Raise hand and unmute Mic
- Type in Chat

- Tutorial (~ 1.5 hours)
 - Part-1: Facial Expression Recognition
 - Part-2: Facial Action Units
 - Part-3: Overview of PyTorch
- Break (~ 5 minutes)
- Open question hour (~ 0.5 hour)
 - · Questions, questions, questions...

In case you have a question-

- Raise hand and unmute Mic
- Type in Chat



- Tutorial (~ 1.5 hours)
 - Part-1: Facial Expression Recognition
 - Part-2: Facial Action Units
 - Part-3: Overview of PyTorch
- Break (~ 5 minutes)
- Open question hour (~ 0.5 hour)
 - Questions, questions, questions...

In case you have a question-

- Raise hand and unmute Mic
- Type in Chat

What is an Expression?



the process of making known one's thoughts or feelings

What is an Expression?

What is a Facial Expression?



a look on someone's face that conveys a particular emotion

Humans mostly communicate expressions via face

Positions and movement of facial components express our feeling and emotions

- Often more powerful and expressive in comparison to verbal communication
- Ease of interpretation via human perception

































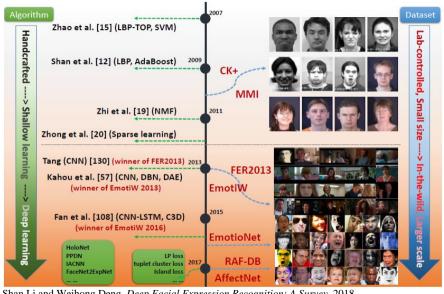
We are interested in automatic perception of expressions





Drum roll...

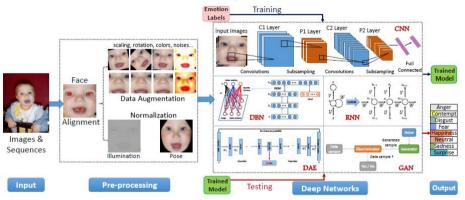
Enter the era of Machine Learning



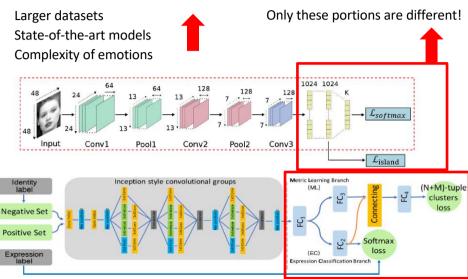
Expression Recognition in 2012...



Expression Recognition in 2020!



What happened?



Underlying feature extraction mechanism remains the same

Shan Li and Weihong Deng, Deep Facial Expression Recognition: A Survey, 2018:

With that said, how do we recognize emotions?

We will look at two methods-

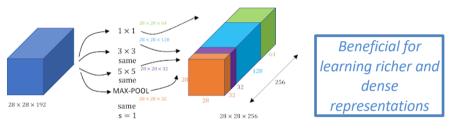
- 1. Deep Separable Layers
- Xception module
- 2. Improvement Techniques



Finetuning Transfer-Learning Class Reweighting

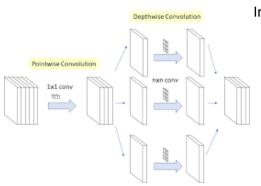
Xception module

Derived from the famous Inception architecture



Split convolutions as per filter sizes

Xception module



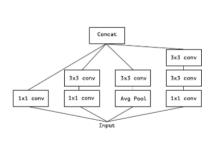
Derived from the famous Inception architecture

Beneficial for learning richer and dense representations

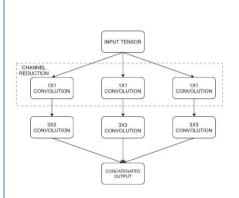
Split convolutions as per channels

Xception module

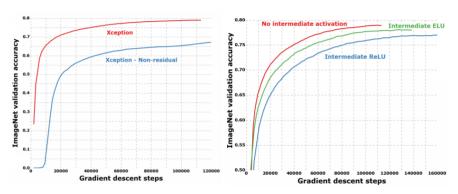
Inception



Xception



Xception module



Xception module

- You will implement Xception for Task-1
- Note that only the layers of the model need to be implemented
- · Keep a track of your loss and accuracy metrics
- Visualize as much as you can!

Improvement Techniques

1. Finetuning

Train only a portion of the model

Source Target model model Random Train from Output layer Output laver initialization scratch Layer L - 1 Laver L - 1 000V · · · · • Prostrain Fine-hine copy Layer 1 copy ----≠ Layer 1 Source data Target data

Avoid training large models from scratch!

- Tips-
 - Slower learning rate
 - Freezing earlier layers
 - Train for less epochs
 - Monitor performance frequently

Improvement Techniques

1. Finetuning

Advantages-

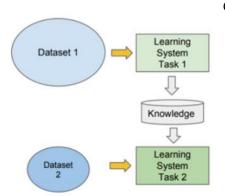
- Works well for very large models
- Scalable to larger datasets

Disadvantages-

- May degrade performance when model not initialized well
- Only applicable to structurally similar datasets

Improvement Techniques

2. Transfer-Learning



Similar to finetuning but for different domains

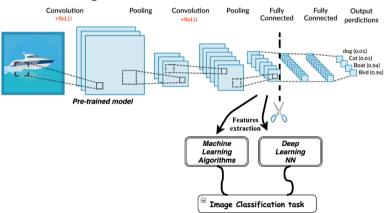
Source task



Target task

Improvement Techniques

2. Transfer-Learning



Improvement Techniques

2. Transfer-Learning

Advantages-

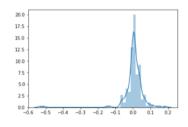
- · Works well for complex datasets
- Knowledge can be transferred to smaller models

Disadvantages-

- Not suitable for different datasets (dog vs cat)
- Model may suffer from catastrophic forgetting

Improvement Techniques

2. Class-Reweighting



Datasets are often biased towards class distributions

Reweight classes as per a criterion-

- Frequency of samples
- Train/Test split
- Application at hand

Just because the model was trained on 10 classes does not mean it will classify all 10 correctly

Improvement Techniques

2. Class-Reweighting

Advantages-

- Addresses dataset bias
- Only trains the model on essential data samples

Disadvantages-

- May not be effective when all classes are extremely large
- Difficult to come up with a practical weighting scheme

- Tutorial (~ 1.5 hours)
 - Part-1: Facial Expression Recognition
 - Part-2: Facial Action Units
 - Part-3: Overview of PyTorch
- Break (~ 5 minutes)
- Open question hour (~ 0.5 hour)
 - · Questions, questions, questions...

- In case you have a question-
- Raise hand and unmute Mic
- Type in Chat

Often difficult to naively interpret emotions





Which of the above two is sad? Why?

Expressions can have different semantic meaning



Sad

Bored

Lazy

Tired

Disappointed

A face in general does not convey a lot of information when compared to its individual components

We as humans already know which portions of the face indicate a particular emotion







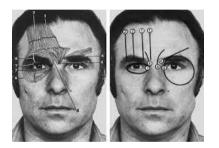




Sad? Why?

Happy? Why?

Ekman's Theory



Certain regions and movements of the face characterize emotions

Ekman's Theory



Group each such unit by its action on the face, hence Facial Action Unit (FAU)

Ekman's Theory





Turns out there are quite a lot of FAUs, do we need to recognize all of them?!

Ekman's Theory

Main FAUs, other emotions are a combination of these units

6 6	AU 1	-	AU 12
	AU 4	has	AU 14
	AU 6	NEV.	AU 15
3 5	AU 7	No.	AU 17
(As)	AU 10	(e	AU 23

Not really!

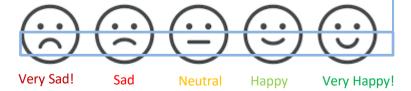
With that said, how do we recognize FAUs?

FAU Recognition

We generally focus on FAU intensity estimation

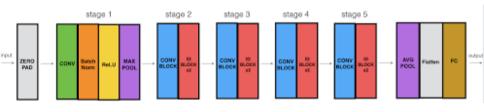
This helps us quantify the degree of emotion

Only need to look at this region, how cool is that!



FAU Recognition

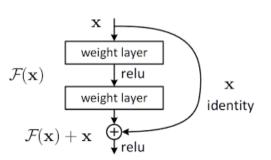
Lets look at a simple and effective method
The ResNet



Kaiming He, Xiangyu Zhang, Shaoqing Ren, and Jian Sun. *Deep residual learning for image recognition*. In Proceedings of the IEEE conference on computer vision and pattern recognition, 2016.

FAU Recognition

Lets look at a simple and effective method
The ResNet



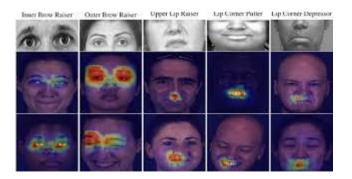
Residual connections help with-

- Richer representations
- Vanishing gradients
- Consistency in local features

Kaiming He, Xiangyu Zhang, Shaoqing Ren, and Jian Sun. Deep residual learning for image recognition. In Proceedings of the IEEE conference on computer vision and pattern recognition, 2016.

FAU Recognition

FAU intensities using deep ResNets



Kaiming He, Xiangyu Zhang, Shaoqing Ren, and Jian Sun. Deep residual learning for image recognition. In Proceedings of the IEEE conference on computer vision and pattern recognition, 2016.

FAU Recognition

You will be implementing and improving ResNet for Task 2

You need to implement the bottleneck block along with deconvolutional layers

Please read the starter code carefully to avoid errors!

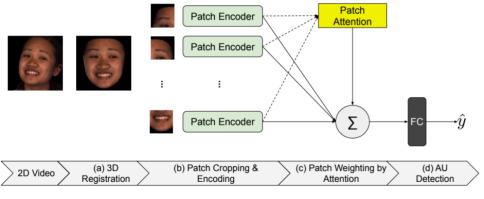
Model is pretrained and will only work if your implementation is correct

Visualize as much as you can!

Can we improve FAU recognition?

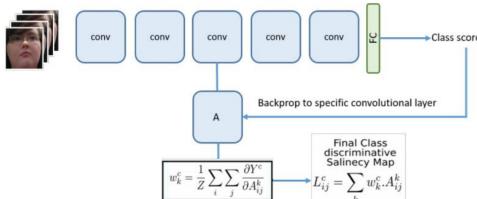
- 1. Paper-1: D-pattnet: Dynamic patch-attentive deep network for action unit detection.
- 2. Paper-2: Attention-based facial behavior analytics in social communication.
- 3. Paper-3: Large margin loss for learning facial movements from pseudoemotions
- 4. Paper-4: Unmasking the devil in the details: What works for deep facial action coding?

Paper-1



Itir Onal Ertugrul, Le Yang, Laszlo A Jeni, and Jerey F Cohn. D-pattnet: *Dynamic patch-attentive deep* network for action unit detection. Frontiers in computer science, 1:11; 2019.

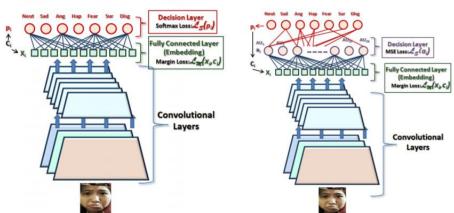
Paper-2



Lezi Wang, Chongyang Bai, Maksim Bolonkin, Judee Burgoon, Norah Dunbar, VS Subrahmanian, and Dimitris N Metaxas. *Attention-based facial behavior analytics in social communication*. In 30th BMVC,

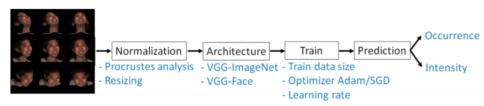
2020.

Paper-3



Learning expressions
Andrei Racoviteanu, Mihai-Sorin Badea, Corneliu Florea, Laura Florea, and Constantin Vertan. Large
margin loss for learning facial movements from pseudo-emotions.

Paper-4



Koichiro Niinuma, Laszlo A. Jeni, Itir• Onal Ertugrul, and Jerey F. Cohn. *Unmasking the devil in the details: What works for deep facial action coding?* In 30th British Machine Vision Conference 2019.

Today

- Tutorial (~ 1.5 hours)
 - Part-1: Facial Expression Recognition
 - Part-2: Facial Action Units
 - Part-3: Overview of PyTorch
- Break (~ 5 minutes)
- Open question hour (~ 0.5 hour)
 - · Questions, questions, questions...

- In case you have a question-
- Raise hand and unmute Mic
- Type in Chat

How do we train a model?

$$\theta^* = \arg\min_{\theta} \sum_{(x,y) \in \mathcal{D}} ||f_{\theta}(x) - y||_2^2$$

$$\theta^* = \min_{\theta} \sum_{(x,y)\in\mathcal{D}} ||f_{\theta}(x) - y||_2^2$$

gradient descent

neural network

What is PyTorch?

Python library for-

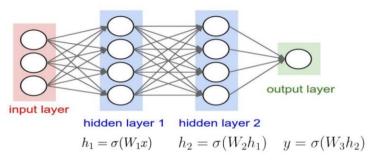
- Defining neural networks
- · Automatically computing gradients

PyTorch alternatives-

- Tensorflow
- JAX
- Chainer
- ...

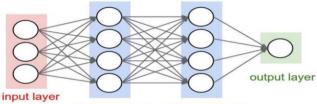
Essentially all do the same thing

How does PyTorch work?



Forward Pass

How does PyTorch work?



hidden layer 1 hidden layer 2

You define	$h_1 = \sigma(W_1 x)$	$h_2 = \sigma(W_2 h_1)$	$y = \sigma(W_3 h_2)$
PT computes	$\frac{\partial y}{\partial W_1} = \frac{\partial y}{\partial h_2} \frac{\partial h_2}{\partial h_1} \frac{\partial h_1}{\partial W_1}$	$\frac{\partial y}{\partial W_2} = \frac{\partial y}{\partial h_2} \frac{\partial h_2}{\partial W_2}$	$\frac{\partial y}{\partial W_3}$

[picture from Stanford's CS231n]

Backward Pass

PyTorch Tutorial in Colab

Link also provided in Project-B handout so that you can go through it at your own pace

https://colab.research.google.com/drive/1r4omUxckDpfhMs_py1y WWPprWy7R8rzj?usp=sharing

Today

- Tutorial (~ 1.5 hours)
 - Part-1: Facial Expression Recognition
 - Part-2: Facial Action Units
 - Part-3: Overview of PyTorch
- Break (~ 5 minutes)
- Open question hour (~ 0.5 hour)
 - Questions, questions, questions...

In case you have a question-

- Raise hand and unmute Mic
- Type in Chat

Today

- Tutorial (~ 1.5 hours)
 - Part-1: Facial Expression Recognition
 - Part-2: Facial Action Units
 - Part-3: Overview of PyTorch
- Break (~ 5 minutes)
- Open question hour (~ 0.5 hour)
 - Questions, questions, questions...

In case you have a question-

- Raise hand and unmute Mic
- Type in Chat

Open Question Hour

Questions...

In case you have a question-

- Raise hand and unmute Mic
- Type in Chat