

# MSc Project 2021

Title: Estimating personality in communication

Name: Yuichi Midorikawa

weekX This indicates when it was done

X-X This corresponds to the mindmap number

Time Plan

Task

1 what they say.

analysis

1-1. Find a dataset to use

2-1. Find a model to use

3-1. Find a model to use

4-1. Find a model to use

capture a dataset that contains people talking and the text of

using the conversation text, do sentiment analysis (A)

2-2. Using the model and its data set, perform sentiment

3-2. Using the model and its data set, extract body pose

4-2. Using the model and its data set, extract facial points

5-3. Predict (A) from (B) body pose + (C) facial feature points

3 from the videos, extract people and body pose (B)

4 from the head, extract facial feature points (C)

Then train a model to predict (A) from (B)+(C)

5-2. Predict (A) from (C) facial feature points

5-1. Predict (A) from (B) body pose

evaluate and analyse the results.

6-1. Decide a evaluation metrics

Write a paper

6-2. evaluate and analyse the results

Step

Oct

1-1

2-1

3-1

4-1

2-2

3-2

4-2

5

5-1

5-2

5-3

6

6-1

6-2

11

18

Nov

8

buffer

15

Improving the accuracy of the

model!

25

|week2|week3|week4|week5|week6|week7

Dec

buffer

29

week8 week9 week10

My Progress

Status

Finished

Finished Finished

Finished

Finished

Finished

Finished

Finished

Finished

Finished

Finished

Runnina

Running

Finished

Running Running

Running

Running

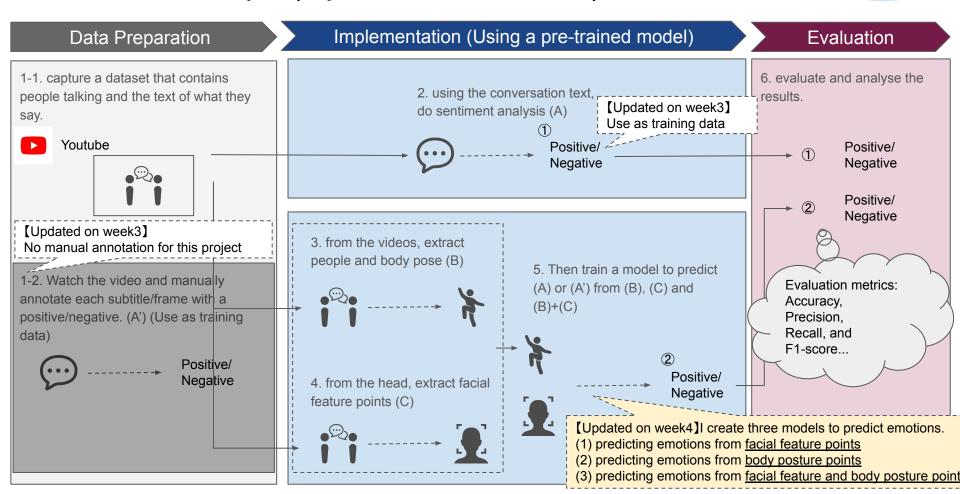
week5

No)

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### Research Steps (Updated on week 5)







## 1. Summary of actions agreed during last meeting

- 1-1. Performed sentiment analysis on each sentence of youtube videos
- 1-2. Extracted facial feature points from the head
- 1-3. Organized the analysis procedure

## 1-1. Perform sentiment analysis on each sentence of a youtube video (Reference) Youtube videos to use for this project

week5

			Number of	neutral		joy s		surprise		anger		sadness		fear		disgust	
	Title	Category	sentences	#N	%	#N	%	#N	%	#N	%	#N	%	#N	%	#N	%
	Why I Don't Use A Smart Phone   Ann Makosinski				61.72		24.88										
1	<u>TEDxTeen</u>	TED	209	129	%	52	%	17	8.13%	6	2.87%	4	1.91%	1	0.48%	0	0.009
					63.60		17.98										
2	How to speak so that people want to listen   Julian Treasure	TED	228	145	%	41	%	16	7.02%	17	7.46%	7	3.07%	0	0.00%	2	0.889
					71.11		13.09										
3	How to spot a liar   Pamela Meyer	TED	405	288	%	53	%	18	4.44%	32	7.90%	12	2.96%	2	0.49%	0	0.009
	Robert Waldinger: What makes a good life? Lessons from				69.81		15.09										
4	the longest study on happiness   TED	TED	212	148	%	32	%	18	8.49%	1	0.47%	12	5.66%	1	0.47%	0	0.009
					62.68		25.00										
5	How to make stress your friend   Kelly McGonigal	TED	284	178	%	71	%	20	7.04%	6	2.11%	5	1.76%	4	1.41%	0	0.009
	How to Get Your Brain to Focus   Chris Bailey				74.18		13.40										
6	<u>TEDxManchester</u>	TED	306	227	%	41	%	18	5.88%	11	3.59%	7	2.29%	0	0.00%	2	0.659
				130	71.43	21	11.54	10		11		7		1		2	
7	Funniest Moss Moments   The IT Crowd   Part 1	Drama	182		%		%		5.49%		6.04%		3.85%		0.55%		1.109
	10 Hilarious Comedy Bits of Series 10   Live at the Apollo			391	74.19	66	12.52	19		28		12		10		1	
8	BBC Comedy Greats	Comedy	527		%		%		3.61%		5.31%		2.28%		1.90%		0.199
	That Song In Every Musical That No One Likes - Sarah			46	75.41	8	13.11	5		2		0		0		0	
9	Smallwood Parsons	Musical	61		%		%		8.20%		3.28%		0.00%		0.00%		0.009
	Michael McIntyre on Google Earth   Michael McIntyre's			42	75.00	9	16.07	1		4		0		0		0	
10	Comedy Roadshow   BBC Comedy Greats	Comedy	56		%		%		1.79%		7.14%		0.00%		0.00%		0.009
					69.80		15.95										
			2470	1724	%	394	%	142	5.75%	118	4.78%	66	2.67%	19	0.77%	7	0.289

This sentiment analysis is based on BERT's pre-trained model (GoEmotions).

https://github.com/monologg/GoEmotions-pytorch GoEmotions: A Dataset of Fine-Grained Emotions https://arxiv.org/pdf/2005.00547.pdf

In every video, 'neutral' accounted for more than 60%.

The emotions of 'sadness', 'fear' and 'disgust' accounted for less than 5%.

## 2-1. Perform sentiment analysis on each sentence of a youtube video (Reference) Youtube videos to use for this project

week5

	(* ************************************					j `											
			Number of	umber of neutral		joy		surprise		anger		sadness		fear		disgust	
	Title	Category	sentences	#N	%	#N	%	#N	%	#N	%	#N	%	#N	%	#N	%
	Why I Don't Use A Smart Phone [ Ann Makosinski ]				61.72		24.88										
1	<u>TEDxTeen</u>	TED	209	129	%	52	%	17	8.13%	6	2.87%	4	1.91%	1	0.48%	0	0.00
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2	How to speak so that people want to listen   Julian Treasure	TED	228	145	%	41	%	16	7.02%	17	7.46%	7	3.07%	0	0.00%	2	0.88
					71.11		13.09										
3	How to spot a liar   Pamela Meyer	TED	405	288	%	53	%	18	4.44%	32	7.90%	12	2.96%	2	0.49%	0	0.00
	Robert Waldinger: What makes a good life? Lessons from				69.81		15.09										
4	the longest study on happiness   TED	TED	212	148	%	32	%	18	8.49%	1	0.47%	12	5.66%	1	0.47%	0	0.00
What's next?					62.68		25.00										
				178	%	71	%	20	7.04%	6	2.11%	5	1.76%	4	1.41%	0	0.00
	I will try to find other videos to balance the				74.18		13.40										
	motions.(especially, sadness, fear and di		306	227	%	41	%	18	5.88%		3.59%	7	2.29%	0	0.00%	2	
	I will reduce the number of 'neutral' befor			130	71.43	21	11.54	10		11		7		1		2	l
			182		%		%		5.49%		6.04%		3.85%		0.55%		1.10
	nalysis(training) to balance the emotiona	1		391	74.19	66	12.52	19		28		12		10		1	
C	ategories.		527		%		%		3.61%		5.31%		2.28%		1.90%		0.19
				46	75.41	8	13.11	5		2		0		0		0	l
9	Smallwood Parsons	اد _	61		%		%		8.20%		3.28%		0.00%		0.00%		0.00
	Michael McIntyre on Google Earth   Michael McIntyre's			42	75.00	9	16.07	1		4		0		0		0	l
10	Comedy Roadshow   BBC Comedy Greats	Comedy	56		%		%		1.79%		7.14%		0.00%		0.00%		0.00
					69.80		15.95					I					
			2470	1724	%	394	%	142	5.75%	118	4.78%	66	2.67%	19	0.77%	7	0.289

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This sentiment analysis is based on BERT's pre-trained model (GoEmotions).

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The emotions of sadness, fear and disgust accounted for less than 5%.



### 2-2. from the head, extract facial feature points

I will use <u>OpenFace</u> which has close to human accuracy and allows me to get the facial feature 128-dimensional representation.

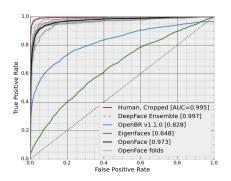


Figure 8: ROC curve on the LFW benchmark with area under the curve (AUC) values.



OpenFace: A general-purpose face recognition library with mobile applications https://elijah.cs.cmu.edu/DOCS/CMU-CS-16-118.pdf 2-3. Data collection procedure and modules

step	Action item	Manual/Automatic	Modules (ipynb, sh) week5
1	Find a Youtube video	Manual	-
2	Download the video	Automatic	01_youtube_downloader.ipynb
3	Get subtitles for the video and paste them into Excel	Manual	-
4	Organise start times and sentences on one line in Excel	Automatic	02 organized excel file.ipynb
5	Find the end time and the difference (end time - start time)	Automatic	03_adding_time_information.ipynb
6	Looking for emotion in every sentence	Automatic	04_emotion_extraction_from_subtitles.ipynb
7	Cut out videos (mp4, avi) using playing time and differences	Automatic	05_ffmpeg.ipynb
8	Extract facial feature points from the head using OpenFace	Automatic	06_run_openface.ipynb, run_openface.sh
9	Data pre-processing	Automatic	To be updated
10	Merge csv files output by OpenFace	Automatic	08_openface_result_merged.ipynb
11	Extract body pose points from the head using OpenPose	Automatic	To be updated
12	Data pre-processing	Automatic	To be updated
13	Merge csv files output by OpenPace	Automatic	To be updated
14	Merge csv files output by OpenFace and OpenPace	Automatic	To be updated



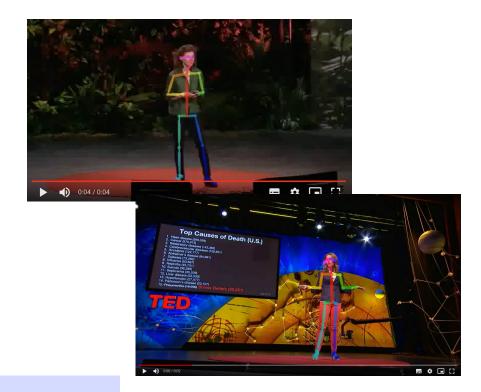
## 2. Summary of work done & results this week

- 2-1. Extracted body pose feature points from the body using OpenPose
- 2-2. Created the model and predicted emotions from facial features.
- 2-3. Dissertation: I have written a draft version of Chapter 1 (Introduction part).
  - → we will look at 'Time plan for writing my dissertation'



## 2-1. Extracted body pose feature points from the body

I will use OpenPose which has represented the first real-time multi-person system to jointly detect human body, hand, facial, and foot keypoints (in total 135 keypoints) on single images.





### 2-1. Extracted body pose feature points from the body

Output example

https://drive.google.com/drive/folders/1wZNmikjedBjwPyTCUARVbabgZhJq0fMS

#### 2-2. Created the model and predicted emotions from facial features.

I predicted emotions from facial features using LogisticRegression().

Model	accuracy
LogisticRegression()	0.6858

#### What's next?

- 1. I will try to use other models (<u>RandomForestClassifier()</u>, <u>GradientBoostingClassifier()</u>, <u>AdaBoostClassifier()</u>, <u>SVC()</u>, <u>LinearSVC()</u>) to get the accuracy. + NN
- 2. I will do feature selection.
- 3. I will do data pre-processing.

#### Time plan for writing my dissertation

		Nov		Dec	My Progres			
What sh	ould I write?	1	8	15	22	29	6	
Level 1	Level 2	week5	week6	week7	week8	week9	week10	Status
1 Introdu	oction						Buffer	Draft
	Briefly explain the context of the project problem							Draft
	Specify overall aim and objectives and report structure							Draft
2 Analysis	s/ Requirements							Draft
	Problem Statement							Draft
	Background Survey/Analysis							Draft
	Effectively combine above in one chapter							Draft
3 Design &	& Implementation							-
	Discuss the main features of your design and how it evolved							-
	In your implementation part							-
4 Testing8	Evaluation							-
	Describe how you evaluated your solution/product							-
	Summarise the evaluation results, and use them to critically evaluate your own work							-
	Be honest about any shortcomings							-
5 Conclus	ion							-
	Describe the status of your research/product							-
	Summarize what you have achieved							-
	Compare to what you originally set out to achieve							-
	Relate your work to relevant previous work							-
	Suggest further/future work that you think would be worthwhile							-
6 Bibliogra	aphy							-
	List, in alphabetical order by author and date, all articles that you have consulted							-
	Use consistent style							-
	Collect all the details when you access a document first							-

- 3. Questions to be discussed during the meeting
- 3-1. Is it better to predict emotions on a frame-by-frame basis? Or should it be done on a sentence(= several frames)-by-sentence basis?
- → In the future, I would like to predict emotion based on facial expressions and posture features in a sentence.

if frame-by-frame

- 3-2a. What kind of model is available for sentence-by-sentence prediction?
- → LSTM, Transformer

if sentence-by-sentence

- 3-2b. How is the data pre-processing for frame-by-frame prediction?
- → read research paper

#### 4. Proposed objectives for next week

#### 4-1. I will visualize data for feature selection

e.g.

- 1. Find features that are correlated with emotion (heat maps)
- 2. Get the variance and standard deviation for each emotion

#### 4-2. I will conduct data pre-processing to facial feature points.

e.g.

- 1. Reduce the bias of the emotional categories (less neutral, more sadness)
- 2. Turn the face of the video to the front (turn a sideways face to the front)
- 3. Calculate the amount of change (movement) of the face coordinates and add it to the feature set.

#### 4-3. Dissertation: Chapter 2 and Chapter 3



#### 5. Articles read this week

5-1. Analysis of Emotion Recognition using Facial Expressions, Speech and Multimodal Information

https://dl.acm.org/doi/10.1145/1027933.1027968

## End