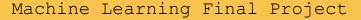


# THAI TEXT EMUTUN DETECTION



















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01

**OBJECTIVE** 



## **Objectives**







- Further analysis in other related field such as marketing
- For understanding emotion in thai language



















#### STEP 1



- $\bullet$  0 = SAD
- 1 = JOY AND HAPPINESS
- $\bullet$  2 = LOVE
- $\bullet$  3 = ANGER
- $\bullet$  4 = FEAR
- 5 = SURPRISE

 We create the data by gathering data from x.com

 Ask an opinion on our friends in each sentence



## **IMPORT**



STEP 2

• Save excel as csv.

• Export the saved file to Github.

 Use google collab to import the saved data from Github.



### **EXPLORE**

• Check for error and missing values.

Check data size.

#### STEP 3







#### **PREPROCESS**

Import PYTHAINLP library.

Change emoji to thai words using stopword from THAINLP library.

• Tokenize thai words - to move all the emoji to the back and eliminating the meaningless word as well as split data into words.

₁



#### STEP 5



#### **MODEL**

 Use countvectorizer to create bag of words for Multinomial NB and Logistic Model.

• Use TF-IDF for Logistic regression.

 Make Multinomial NB and Logistic Model.





### **LIBRARY USED**







#### **SKLEARN**



Provides simple and efficient tools for data analysis and modeling.

#### **PYTHAINLP**



Library used for tokenizing thai language.

#### **PANDAS**



Library used for creating DataFrames for training set.



### LIBRARY USED







#### **MATPLOTLIB YELLOWBRICK**



Helps generate plots and charts for the model.



Library built on top of Matplotlib design to enhance ML model

#### **NUMPY**



Library used for numerical computing in a large multi-dimensional arrays



### LIBRARY USED





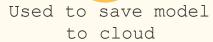


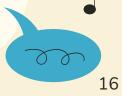
#### **STREAMLIT**



Allows developers to turn data scripts into shareable web apps with minimal code.













## **DATA**

DATA from X.com





## **DATA**

#### **DATA DESCRIPTION**

- $\bullet$  0 = SAD
- 1 = JOY AND HAPPINESS
- 2 = LOVE
- 3 = ANGER
- $\bullet$  4 = FEAR
- 5 = SURPRISE





#### DATA

#### **CRITERIA**

- The data we are choosing is a Data in X.com which is the most frequent community in Thai needs and young adults to comment their thought on which has a unique slag and language
- Hashtag we select hashtag
   according to our target such as #
   เศร้า #รัก and a trendy hashtag such
   as #อุงเอิง #โยเกิร์ต #พีเค

- The period of gathering is a recent 2-3 weeks (25 Feb-6 Mar) (for an update data as a new slay is always created)
- There are 315 rows of data with 6 columns
- We manually label the opinion on each sentence emotion expression by asking our friend to comment of the opinion.



## **DATA from X.com**

	Text	ans1	ans2	ans3	ans4	ans5	Final	Clean_Text
0	ครูอึ้ง! โดนเซอร์ไพรส์ แต่ดันเจอเซอร์ไพรส์กลับ	5	5	5	5	5	5	ครูอึ้ง! โดนเซอร์ไพรส์ แต่ดันเจอเซอร์ไพรส์กลับ
1	วันนี้พระจันทร์น่ารักมาก :-)	1	2	2	2	1	2	วันนี้พระจันทร์น่ารักมาก :-)
2	ฮือฮา! 'พระพุทธรูปสิ่งศักดิ์สิทธิ์' ที่อยู่ในส	5	5	5	5	5	5	ฮือฮา! 'พระพุทธรูปสิ่งศักดิ์สิทธิ์' ที่อยู่ในส
3	"กลัวเธอจะหนีไปกลัวเธอจะใจร้าย"	4	4	4	0	4	4	"กลัวเธอจะหนีไปกลัวเธอจะใจร้าย"
4	มีคนแปลกหน้า เข้ามาหาเรื่องเรา	3	3	3	3	3	3	มีคนแปลกหน้า เข้ามาหาเรื่องเรา
310	มันหนึบคือเหี้ยไรว่ะ รสชาสเหมือนรองเท้านักเรีย	3	3	3	0	3	3	มันหนึบคือเหี้ยไรว่ะ รสชาสเหมือนรองเท้านักเรีย
311	มองผ่านเหมือนแมลงสาบบินอ่ะ กัว แง	4	4	4	4	4	4	มองผ่านเหมือนแมลงสาบบินอ่ะ กัว แง
312	รู้เลยบ้านนี้ใครใหญ่สุด กัว	4	4	4	4	4	4	รู้เลยบ้านนี้ใครใหญ่สุด กัว
313	บาร์โค้ดทำถึงเกินอะ กัว 🤯 🤯 🤯	4	4	4	4	4	4	บาร์โค้ดทำถึงเกินอะ กัว 🤯 🤯 🤯
314	ในไอจีมาลงรูปรองเท้าผูกกันเพิ่มด้วย กัว อะ 😭 🤯 😭	4	4	4	4	4	4	ในไอจีมาลงรูปรองเท้าผูกกันเพิ่มด้วย กัว อะ 🤯 🤯 🤯





## **WHY THAI TEXT**

Thai sentence is not straightforward







## **MODEL**



## **Testing and Training dataset**

XX

x\_train,x\_test,y\_train,y\_test = train\_test\_split(X,y,test\_size=0.2,shuffle=True, stratify = y,random\_state=2002)





 We would like to training the dataset as much as possible so we set the training dataset to be 80% with shuffle and stratify the dataset

## **MODELS**





#### **Multinomial NB**

There are many category output not as binomial and calculate from probability with no requirement of features



## **Logistic Regression**

It is
classification
that can find
probability and
can do multinomial
with no
requirement of
feature.

## **MODELS**







#### **Multinomial NB**

There are many category output not as binomial.

#### Step:

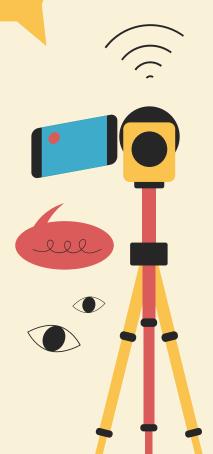
- 1. We tokenized the data using THAINLP library
- Splitting data into testing and training dataset
- 3. Using countvectorizer to count the frequency of the words
- 4. Then create the model.



#### Why use multinomial NB instead of other classification model?

Multinomial NB could handle the multiple class classification since there are 6 targets in our model

- $\bullet$  0 = SAD
- 1 = JOY AND HAPPINESS
- $\bullet$  2 = LOVE
- 3 = ANGER
- 4 = FEAR
- 5 = SURPRISE



#### **MODELS**

#### Step:

- 1. We tokenized the data using THAINLP library
- Splitting data into testing and training dataset
- 3. Using countvectorizer to count the frequency of the words
- 4. Use TF-IDF to eliminate the frequent word appeared
- 5. Then create the model.





It is classification that can find probability and can do multinomial.

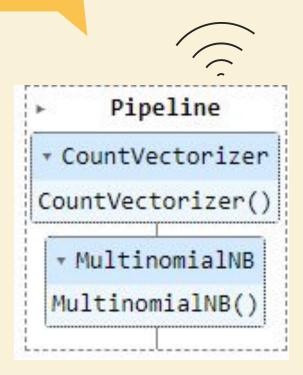




#### **PIPELINE**

- Automate the ML workflow
- Analyze the complexity of language without caring the coefficient
- Allow the creator to focus on the updated data only
- Reduce Model Error





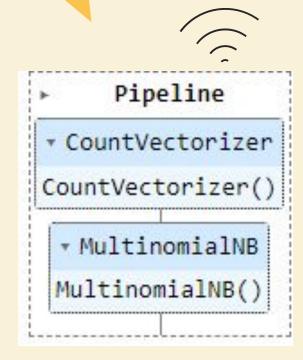


#### **CountVectorizer**

CountVectorizer is a bag of word that we use to count the word

#### Why using countvectorizer instead of other vectorizer?

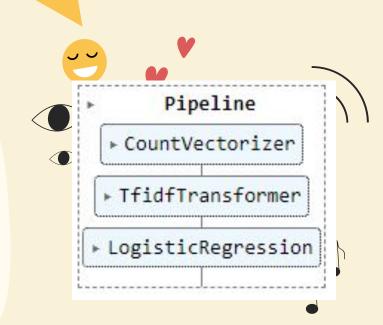
Basic and common use vectorizer
 that help the ML to better
 understanding the model



#### **TF-IDF**

We use TF-IDF function as

- it eliminate the frequent word appeared
- Suitable with logistic regression
- Making more understanding and more accuracy in Logistic Regression







## **RESULT**

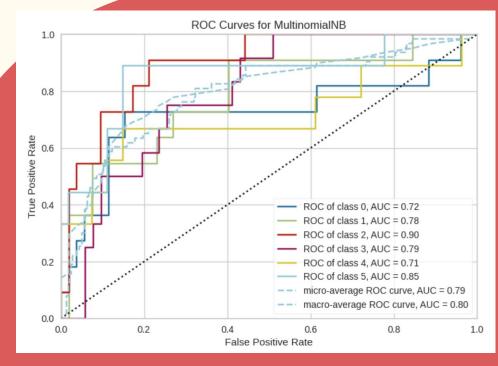


## **Multinomial NB**

Classification	Report:						
	precision	recall	f1-score	support			
Sad	0.53	0.73	0.62	11			
Нарру	0.50	0.27	0.35	11			
Love	0.64	0.64	0.64	11			
Angry	0.42	0.67	0.52	12			
Fear	0.50	0.33	0.40	9			
Surprise	0.67	0.44	0.53	9			
accuracy			0.52	63			
macro avg	0.54	0.51	0.51	63			
weighted avg	0.54	0.52	0.51	63			

F1 score = 0.5090252989683729 Accuracy = 0.5238095238095238 Accuracy on train: 0.968

Accuracy on test: 0.524





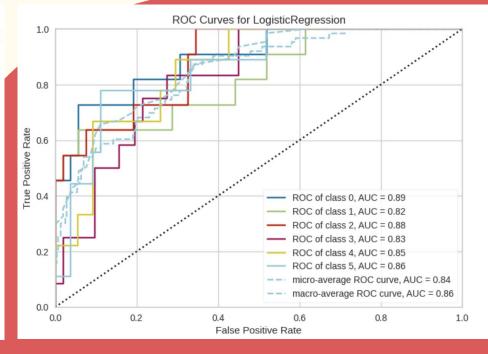
## **Logistic Regression**

Classification	Report:
----------------	---------

	precision	recall	f1-score	support
Sad	0.67	0.55	0.60	11
Нарру	0.50	0.64	0.56	11
Love	0.50	0.64	0.56	11
Angry	0.42	0.67	0.52	12
Fear	0.67	0.22	0.33	9
Surprise	1.00	0.44	0.62	9
accuracy			0.54	63
macro avg	0.63	0.53	0.53	63
weighted avg	0.61	0.54	0.53	63

F1 score = 0.530807830162669 Accuracy = 0.5396825396825397 Accuracy on train: 0.992

Accuracy on test: 0.54



## **Example Testing**



```
ex1 = "แต่หนูโกดไมมหาลัยต้องให้หนูเรียนวันเสาร์ นอยๆๆๆๆ"
x = pipe_lr.predict([ex1])
emotions[x[0]]
'Happy'
ex1 = text process(ex1)
print(ex1)
x = pipe_lr.predict([ex1])
emotions[x[0]]
แต่ หนู โก ดใม มหาลัย ต้อง ให้ หนู เรียน วัน เสาร์ นอ ย
'Angry'
y = pipe_lr.predict_proba([ex1])
array([[0.1823474 , 0.14842353, 0.14532417, 0.32425434, 0.10060137,
        0.09904918]])
x = pipe_nb.predict([ex1])
emotions[x[0]]
'Angry'
```













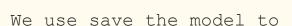


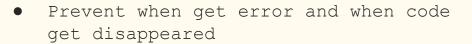




### **Save Model**







- To make an app for further analyze of our model
- from google.colab import drive
  drive.mount('/content/drive')
- #Saving the model to run app import joblib joblib.dump(pipe\_lr, open('/content/drive/MyDrive/ML Project/emotion\_model.pkl','wb'))





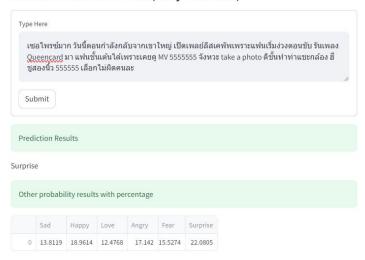




## **Web Application**

#### **Thai Text Emotion Detection**

Detect Emotions In Text (Only Thai text)







"The model shows to be 50%-55% accuracy, but still acceptable"













- Increase data set.
- Average the type of emotion in the data set.





**IMPROVEMENT** 







# THANK







