

# Reconhecimento de Emoções em Tweets

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# **Objetivo**

• Realizar uma análise comparativa entre diferentes abordagens de aprendizado de máquina para identificação de emoções.

- MACHINE LEARNING
  - Logistic Regression
  - Multinomial NB
  - KNN
  - Decision tree
  - Random Forest

- DEEP LEARNING
  - Modelo com Embedding
  - CNN
  - LSTM



# **Repositório Corpus**

sadness (0), joy (1), love (2), anger (3), fear (4), surprise (5)

text (string)	label (class label)
"i didnt feel humiliated"	0 (sadness)
"i can go from feeling so hopeless to so damned hopeful just from being around someone who cares and is awake"	0 (sadness)
"im grabbing a minute to post i feel greedy wrong"	3 (anger)
"i am ever feeling nostalgic about the fireplace i will know that it is still on the property"	2 (love)
"i am feeling grouchy"	3 (anger)
"ive been feeling a little burdened lately wasnt sure why that was"	0 (sadness)
"ive been taking or milligrams or times recommended amount and ive fallen asleep a lot faster but i also feel like so funny"	5 (surprise)
"i feel as confused about life as a teenager or as jaded as a year old man"	4 (fear)
"i have been with petronas for years i feel that petronas has performed well and made a huge profit"	<b>1</b> (joy)





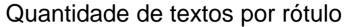
# **Repositório Corpus**

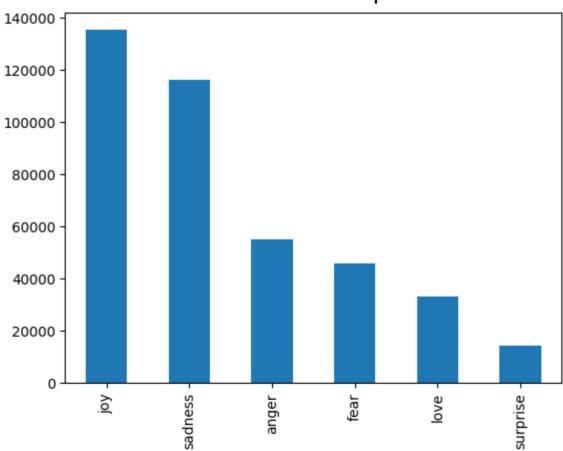
name	train	validation	test
split	16000	2000	2000
unsplit	416809	n/a	n/a

https://huggingface.co/datasets/dair-ai/emotion



# **Repositório Corpus**







# **Trabalhos Relacionados**



#### **Processamento do texto**

- en\_core\_news\_sm
- Tradução de emoji
- Lematização
- Normalização de acentos
- Substituição de caracteres especiais



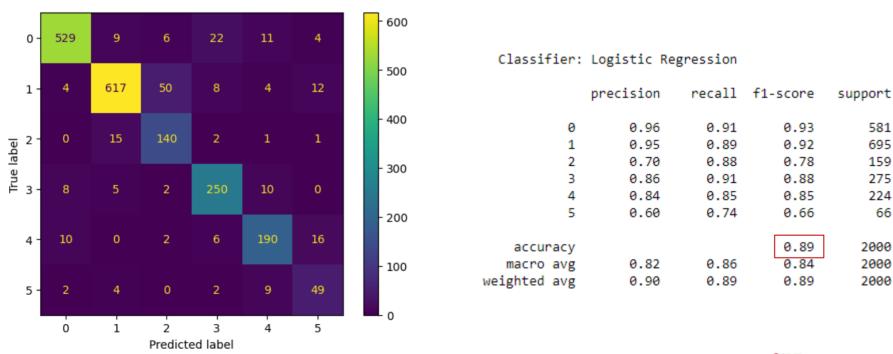
# Vetorização

• CountVectorizer(binary=True, max\_features=None, ngram\_range=(1, 1))

• Vocabulário: 15700 palavras

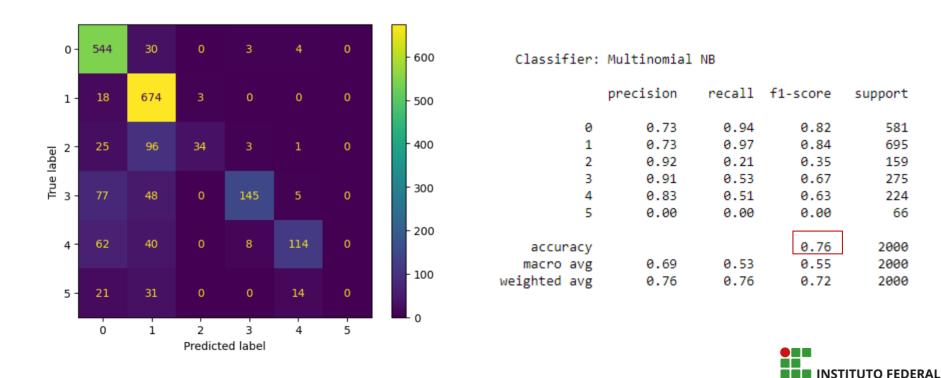


## Logistic Regression



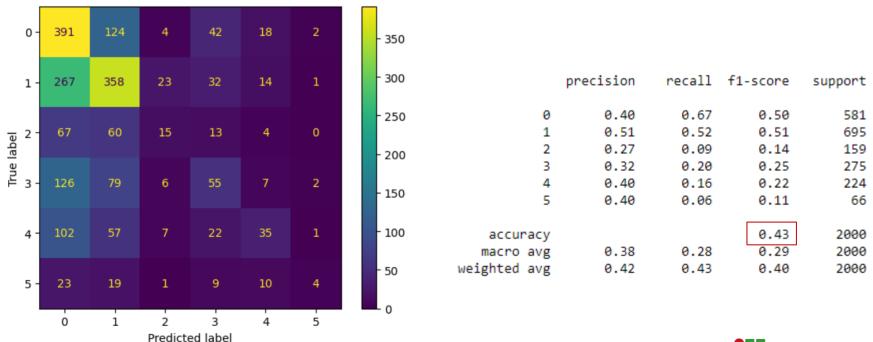


#### Multinomial NB



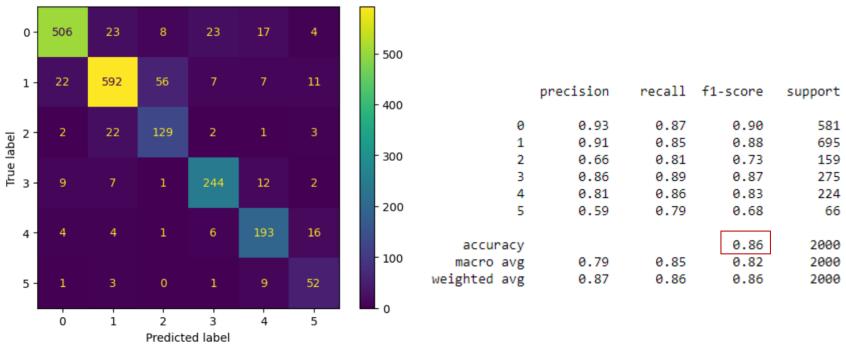
Espírito Santo

#### **KNN**



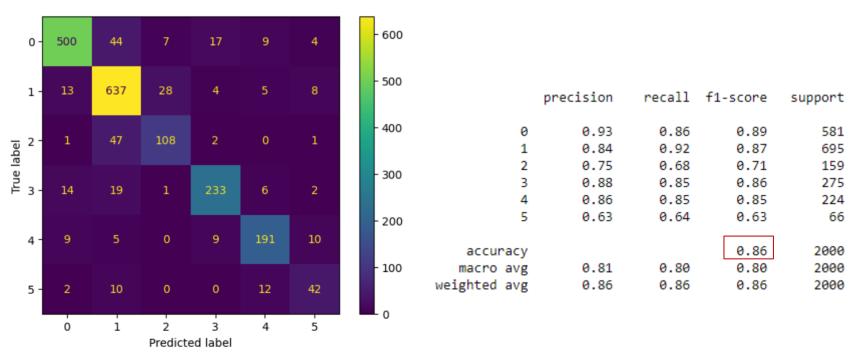


#### **Decision Tree**





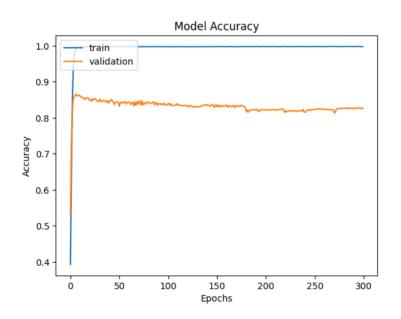
#### Random Forest

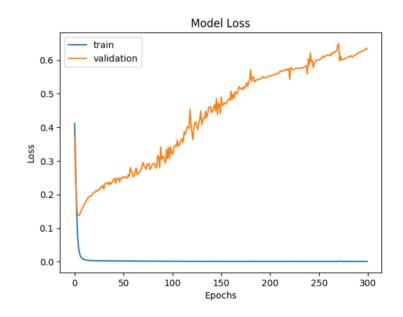




# Modelo simples com Embedding

1,948,654 parâmetros 300 épocas

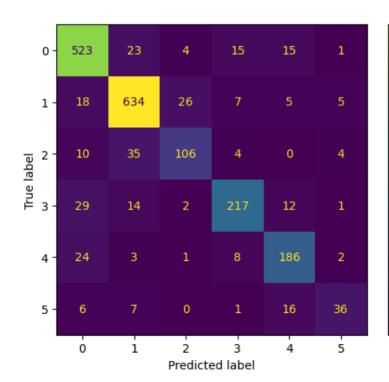


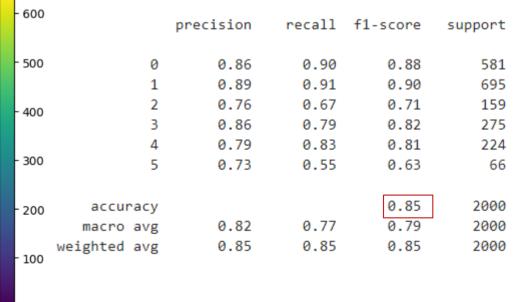




# Modelo simples com Embedding

1,948,654 parâmetros 300 epochs

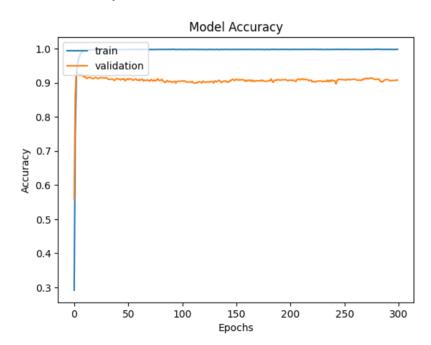


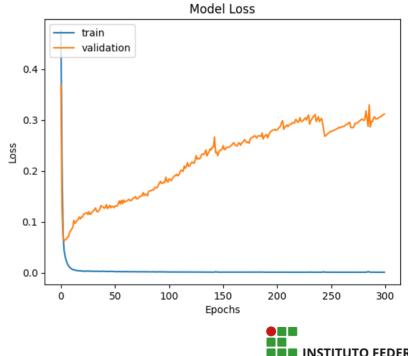




#### **CNN**

1,582,734 parâmetros 300 épocas

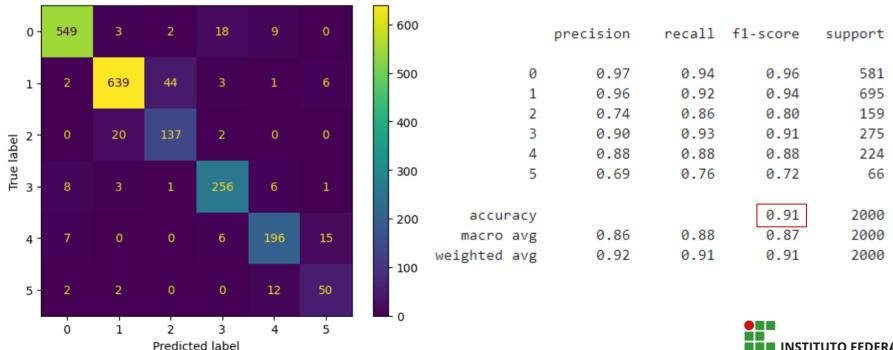






**CNN** 

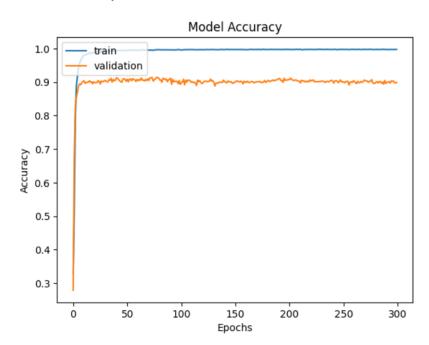
1,582,734 parâmetros 300 épocas

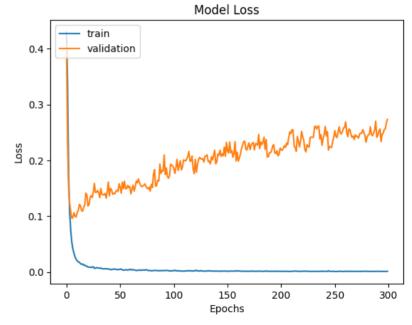




#### **LSTM**

1,821,934 parâmetros 300 épocas

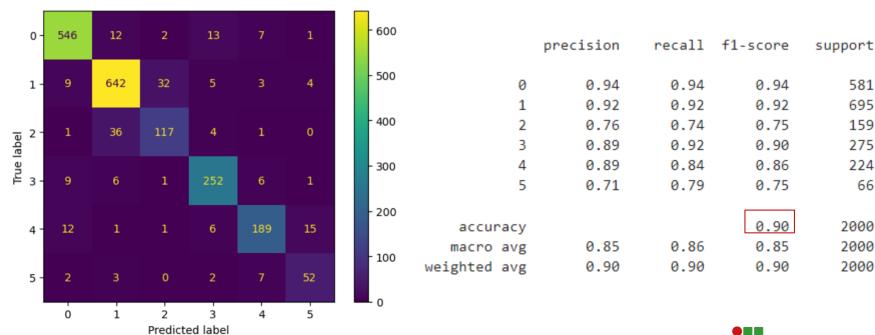






#### **LSTM**

1,821,934 parâmetros 300 épocas





# Comparação entre acurácia do teste de cada técnica

	TÉCNICA	ACURÁCIA		
MACHINE LEARNING	Logistic Regression	0,89		
	Multinomial NB	0,76		
	KNN	0,43		
	Decision Tree	0,86		
	Random Forest	0,86		
DEEP LEARNING	Modelo com Embedding	0,85		
	CNN	0,91		
	LSTM	0,90		



# Comparação entre resultados das técnicas por classe

	Logistic Regression	Multinomial NB	KNN	Decision Tree	Random Forest	Modelo com Embedding	CNN	LSTM
sadness (0)	529	544	391	506	500	532	549	546
joy (1)	617	674	358	592	637	634	639	642
love (2)	140	34	15	129	108	106	137	117
anger (3)	250	145	55	244	233	217	256	252
fear (4)	190	114	35	193	191	186	196	189
surprise (5)	49	0	4	52	42	36	50	52





Educação pública, gratuita e de qualidade