Introduction

In this document, I'll walk through my findings from implementing a GRU and LSTM with various different types of embeddings. Please see the main.ipynb file for my implementation.

To recap, I implemented two types of models, Gated Recurrent Units (GRUs) and Long Short-Term Memory networks (LSTMs). I also implemented two main types of embeddings, a word2vec embedding and a BERT embedding (in four different languages).

The dataset I used contained spam and ham (not spam) text in four languages: English, Hindi, French, and German. My goal was to train the models to most accurately predict the language of the text.

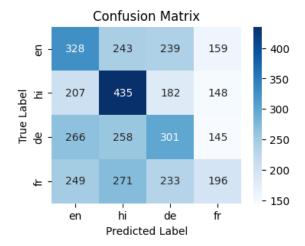
Discussion

See the bottom of the discussion for a table with all results. It was too long to put it before the discussion.

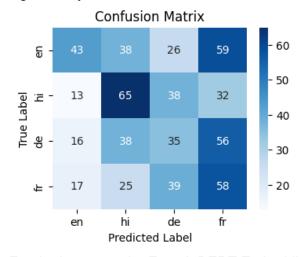
Overall, I found that both the GRU and the LSTM models did a very good job at classifying the entire dataset and the ham dataset when combined with the Word2Vec embeddings. Interestingly, the GRU model did a significantly worse job at differentiating between English and French spam text. In comparison to the LSTM, the GRU model had an accuracy of 0.79 in comparison to the LSTM's accuracy of 0.96. This surprised me because I had read online that GRUs sometimes did better with smaller datasets. LSTMs, however, probably do a better job at understanding the longer dependencies required to distinguish between the two languages.

What was most interesting was how drastically BERT embeddings trained in different languages changed the results. The English BERT embeddings did terribly across the board, with accuracies for GRU and LSTM at around 0.4. However, the English BERT embeddings did clearly distinguish between Hindi and non-Hindi languages. This isn't super surprising, but stresses the importance of having relevant training data.

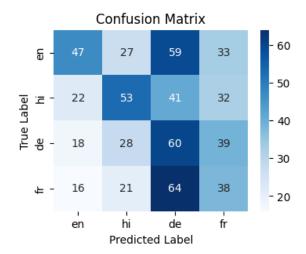
The French BERT Embeddings did even worse, with accuracies around 0.3. What was most interesting was how badly they predicted French language, specifically. Take this confusion matrix from French BERT Embeddings and GRU for ham for example.



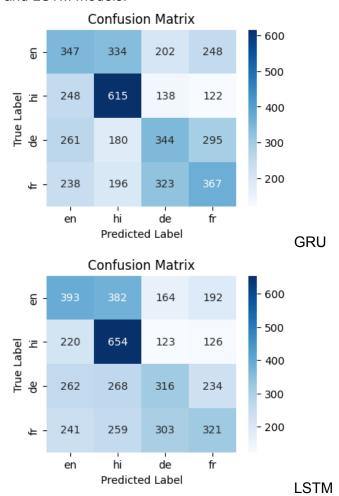
Interestingly, the confusion matrix from French BERT Embeddings and GRU for spam were significantly different.



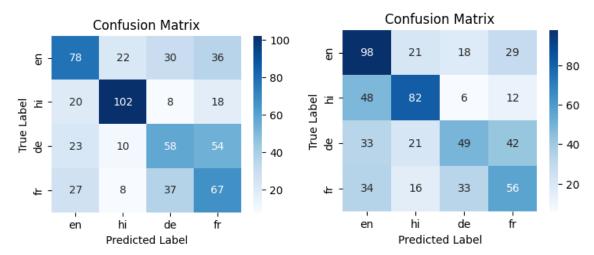
For the ham text, the French BERT Embeddings and GRU seemed almost biased against the French language, whereas, for the spam text, the French BERT Embeddings and GRU did a significantly better job. Interestingly, the LSTM model for the French BERT embedding led similar results, but was biased towards the German language.



Unsurprisingly, the Hindi BERT Embeddings did best at recognizing Hindi text, for both the GRU and LSTM models.



It was interesting that both models did relatively well recognizing the spam text. This may simply be due to the smaller testing size.



Finally, it was interesting to find that the German BERT Embeddings with both the GRU and LSTM did the best across the board for the language-specific BERT Embeddings. The embeddings had between 0.43 and 0.49 accuracy. While this was significantly worse than the Word2Vec embeddings, they were by far higher than the other BERT Embeddings. My hypothesis for why this is the case is that, while French and English are relatively similar synaptically and semantic words are similar, German's grammar structure is dramatically different. While it is still very different from Hindi, it sits at more of a middle ground and thus could classify the spam and ham text relatively better.

Results

Dataset	Embeddings	Model	Confus	sion M	atrix				Accuracy																						
Entire Dataset	Word2Vec	GRU		C	Confusio	n Matrix		_	Accuracy: 0.97397935																						
			- 등	1107	7	12	5	- 1000	Precision: 0.99268514																						
			abel hi	5	1114	2	2	- 800 - 600	Recall: 0.97896004																						
			True Label de hi	35	5	1037	3	- 400																							
			↓ = -	30	6	4	1084	- 200																							
				en	hi Predicte	de d Label	fr																								
Entire Dataset	Word2Vec	LSTM		(Accuracy: 0.9641095																										
				eu -	1039	7	76	76 9 - 100	- 1000	Precision: 0.97292525 Recall:																					
			abel	. 0	1116	6	1	- 800 - 600	0.993688																						
			True Label de hi	9	5	1062	4	- 400																							
			4-	12	9	22	1081	- 200																							
				en	hi Predict	de ed Label	fr	- 0																							
Ham	Word2Vec	GRU			Confusi	on Matr	ix		Accuracy: 0.9619171																						
			윤 -	885	9	62	13	- 800	Precision: 0.97159284 Recall:																						
																									abel hi	- 2	966	1	3	- 600	0.99377376
			True Label de hi	- 8	13	941	8	- 400																							
			- ⊒	- 8	4	16	921	- 200																							
				en	hi Predict	de ted Label	fr																								

Ham	Word2Vec	LSTM		(Confusio	n Matrix	(Accuracy: 0.9715026
			e -	940	5	16	8	- 800	Precision: 0.9898849 Recall:
			True Label de hi '	3	966	0	3	- 600	0.9816672
			True de	28	11	925	6	- 400	
				22	1	7	919	- 200	
				en	hi Predicte	de ed Label	fr	- 0	
Spam	Word2Vec	GRU		(Confusio	n Matrix	(Accuracy: 0.79765886
			e -	70	0	95	1	- 140 - 120	Precision: 0.8181818 Recall: 1.0
			True Label de hi	0	135	9	4	- 100 - 80	
			True l	0	0	143	2	- 60 - 40	
				0	0	10	129	- 20	
				en	hi Predicte	de ed Label	fr	- 0	
Spam	Word2Vec	LSTM		(Confusio	n Matrix	(_	Accuracy: 0.9598662
			- e	157	0	7	2	- 140 - 120	Precision: 0.97949886 Recall: 0.9953704
			True Label de hi '	0	140	4	4	- 100 - 80	0.9953704
			True de	2	0	141	2	- 60 - 40	
			-	0	0	3	136	- 20	
				en	hi Predicte	de ed Label	fr	- 0	

Original	English BERT	GRU		(Confusio	n Matrix	x		Accuracy: 0.41453567																		
	DERI		- eu	567	222	165	177	- 700 - 600	Precision: 0.8190568 Recall:																		
			abel hi	159	725	127	112	- 500	0.767358																		
			True Label de hi	279	268	321	212	- 400 - 300																			
			- ₽	336	285	268	235	- 200																			
			'	en	hi Predicte	de ed Label	fr																				
Ham	English BERT	GRU		(Confusio	n Matrix	x	_	Accuracy: 0.4326425																		
	BERT		- eu	437	167	226	139	- 600	Precision: 0.8107435 Recall:																		
			Label hi	ː= - 104 674 106 8	88	- 500 - 400	0.78830856																				
			True Label de hi	240	224	390	116	- 300																			
			- ₩	268	232	280	169	- 200																			
				en	hi Predicte	de ed Label	fr	- 100																			
Spam	English BERT	GRU		(Confusio	n Matrix	x		Accuracy: 0.51839465																		
	BEIXI		eu -	114	10	11	31	- 100	Precision: 0.86901766 Recall:																		
					i																abel hi	30	93	12	13	- 80	0.7986111
			True Label de hi	31	27	53	34	- 60 - 40																			
			fr -	26	26	37	50	- 20																			
				en	hi Predicte	de ed Label	fr																				

Original	English BERT	LSTM		(Confusio	n Matrix	x		Accuracy: 0.39277703													
	DEKI		eu -	415	257	288	171	- 700 - 600	Precision: 0.7942529 Recall:													
			True Label de hi	121	717	189	96	- 500	0.8307785													
			True de	216	287	427	150	- 400 - 300														
			1 -	226	323	383	192	- 200														
				en	hi Predicte	de ed Label	fr	- 100														
Ham	English BERT	LSTM		(Confusio	n Matrix	K		Accuracy: 0.41709843													
	BLIXI		eu -	460	194	196	119	- 600 - 500	Precision: 0.81905437 Recall:													
			abel hi	103	632	119	118	- 400	0.79695606													
			True Label de hi	197	234	363	176	- 300														
			- 1	287	253	254	155	- 200														
				en	hi Predicte	de ed Label	fr															
Spam	English BERT	LSTM		(Confusio	n Matrix	ĸ	_ 100	Accuracy: 0.43645486													
	BLICE		en -	101	16	25	24	- 100 - 80	Precision: 0.8414634 Recall:													
									ı							abel hi	36	85	15	12	- 60	0.7986111
			True Label de hi	24	41	32	48	- 40														
			fr -	27	38	31	43	- 20														
				en	hi Predicte	de ed Label	fr															

Original	Hindi BERT	GRU		(Confusio	n Matri	x		Accuracy: 0.37528038																																				
			등 -	347	334	202	248	- 600 - 500	Precision: 0.7669441 Recall:																																				
			True Label de hi	248	615	138	122	- 400	0.7754734																																				
			True de	261	180	344	295	- 300																																					
			4 -	238	196	323	367	- 200																																					
				en	hi Predicte	de ed Label	fr																																						
Ham	Hindi BERT	GRU		(Confusio	n Matri	x	_	Accuracy: 0.3746114																																				
			e -	325	282	152	210	- 500	Precision: 0.7725185 Recall:																																				
		abel hi	218	580	68	106	- 400	0.75648564																																					
			True Label de hi	259	198	248	265	- 300 - 200																																					
			↓ -		211	218	293	- 100																																					
				en	hi Predicte	de ed Label	fr																																						
Spam	Hindi BERT	GRU		(Confusio	n Matri	х		Accuracy: 0.5100334																																				
			e -	78	22	30	36	- 100 - 80	Precision: 0.80444443 Recall:																																				
				l																																			True Label de hi	- 20	102	8	18	- 60	0.837963
			True	- 23	10	58	54	- 40																																					
			↓	27	8	37	67	- 20																																					
				en	hi Predicte	de ed Label	fr																																						

Original	Hindi BERT	LSTM		(Confusio	n Matrix	x		Accuracy: 0.37774786
			eu -	393	382	164	192	- 600	Precision: 0.77917415 Recall:
			abel hi	220	654	123	126	- 500 - 400	0.7826871
			True Label de hi	262	268	316	234	- 300	
			1 -	241	259	303	321	- 200	
				en	hi Predicte	de ed Label	fr		
Ham	Hindi BERT	LSTM		(Confusio	n Matrix	ĸ	_	Accuracy: 0.37849742
			- e	292	268	255	154	- 500	Precision: 0.77197707 Recall: 0.79280525
			abel hi	184	567	119	102	- 400	
			True Label de hi	219	186	359	206	- 300	
				196	188	322	243	- 200	
				en	hi Predicte	de ed Label	fr		
Spam	Hindi BERT	LSTM		(Confusio	on Matri	ix		Accuracy: 0.47658864
			e -	- 98	21	18	29	- 80	Precision: 0.8233766 Recall: 0.7337963
			True Label de hi	- 48	82	6	12	- 60	0.7337303
			True de	- 33	21	49	42	- 40	
			- ₩	- 34	16	33	56	- 20	
				en	hi Predict	de ed Label	fr		

Original	French BERT	GRU		(Confusio	n Matrix	ĸ		Accuracy: 0.31426647										
	BERT		eu	383	286	254	208	- 400	Precision: 0.77055216 Recall:										
			True Label de hi	237	438	207	241	- 350	0.75503457										
			True de	297	260	281	242	- 300											
			1	281	310	234	299	- 250											
			en	hi Predicte	de ed Label	fr													
Ham	French BERT	GRU		C	Confusio	n Matrix	ĸ	_	Accuracy: 0.32642487										
BLICI			en -	328	243	239	159	- 400	Precision: 0.7718861 Recall: 0.7502594										
			True Label de hi	207	435	182	148	- 350 - 300											
			True de	266	258	301	145	- 250											
			-	249	271	233	196	- 200											
				en	hi Predicte	de ed Label	fr	- 150											
Spam	French BERT	GRU		(Confusio	on Matri	ix		Accuracy: 0.3361204										
					e -	43		26	59	- 60	Precision: 0.7583497 Recall:								
														True Label de hi	- 13	65	38	32	- 50 - 40
				True de	- 16	38	35	56	- 30										
			fr -	- 17	25	39	58	- 20											
				en	hi Predicte	de ed Label	fr												

Original	French BERT	LSTM		(Confusio	n Matrix	(- 450	Accuracy: 0.2949753		
	BEIXI		eu -	330	297	324	180	- 400	Precision: 0.75549453 Recall:		
			abel hi	246	450	251	176	- 350	0.7439134		
			True Label de hi	298	275	344	163	- 300 - 250			
			- 1	308	355	270	191	- 200			
				en	hi Predicte	de ed Label	fr				
Ham	French BERT	LSTM		(Confusio	n Matrix	(_	Accuracy: 0.31113988		
			eu -	332	197	231	209	- 350 - 325	Precision: 0.77119255 Recall:		
			abel hi	207	367	182	216	- 300	0.7426496		
			True Label de hi	295	187	230	258	- 275 - 250			
			-	242	230	205	272	- 225 - 200			
				en	hi Predicte	de ed Label	fr				
Spam	French BERT	LSTM		(Confusio	on Matri	х		Accuracy: 0.33110368		
			e -	47	27	59	33	- 60	Precision: 0.75959593 Recall:		
					abel hi	- 22	53	41	32	- 50	0.8703704
			True Label de hi	- 18	28	60	39	- 40 - 30			
			fr	- 16	21	64	38	- 20			
				en	hi Predict	de ed Label	fr				

Original	German BERT	GRU		(Confusio	x		Accuracy: 0.49349484					
	BEIXI		en -	575	40	266	250	- 800 - 700	Precision: 0.8166227 Recall:				
			abel hi	77	824	144	78	- 600 - 500	0.744214				
			True Label de hi	369	129	427	155	- 400 - 300					
			- 1	405	60	285	374	- 200 - 100					
				en	hi Predicte	de ed Label	fr						
Ham	German BERT	GRU		(Confusio	n Matrix	x		Accuracy: 0.49948186				
DEIXI	DEIXI		eu -	332	30	236	371	- 700 - 600	Precision: 0.7923052 Recall: 0.84053963				
			abel hi	30	730	116	96	- 500 - 400					
			True Label de hi	197	100	420	253	- 300					
			- π	234	47	222	446	- 200 - 100					
				en	hi Predicte	de ed Label	fr						
Spam	German BERT	GRU		(Confusio	on Matri	ix		Accuracy: 0.46822742				
	DEIXI			e -	84	9	36	37	- 80	Precision: 0.7934509 Recall:			
							abel hi	- 20	89	27	12	- 60	0.7291667
						True Label de hi	46	19	48	32	- 40		
			fr -	51	8	21	59	- 20					
				en	hi Predict	de ed Label	fr						

Original	German BERT	LSTM		(Confusio	n Matri	x	_	Accuracy: 0.48048452															
	BEIXI		eu -	516	45	240	330	- 800 - 700	Precision: 0.8040153 Recall:															
			abel hi	55	820	150	98	- 600 - 500	0.75834084															
			True Label de hi	339	128	394	219	- 400 - 300																
			- #	410	75	227	412	- 200 - 100																
				en	hi Predicte	de ed Label	fr																	
Ham	German BERT	LSTM		(Confusio	n Matri	x	_	Accuracy: 0.48678756															
	BLIN		- eu	454	35	169	311	- 700 - 600	Precision: 0.80911785 Recall:															
			abel hi	55	721	96	100	- 500 - 400	0.75510204															
			True Label de hi	280	119	344	227	- 300																
			4 -	373	49	167	360	- 200 - 100																
				en	hi Predicte	de ed Label	fr																	
Spam	German BERT	LSTM		(Confusio	on Matr	ix		Accuracy: 0.43979934															
	BERT										eu -	57	12	40	57	- 80	Precision: 0.75283444 Recall:							
																								abel hi
					True Label de hi	42	25	51	27	- 40														
			fr -	44	10	21	64	- 20																
				en	hi Predict	de ed Label	fr																	