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Presentation on:

Fake news detection using Temporal Ensembling model

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Course: Statistical Learning for Big Data

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Outline

- Introduction
- Motivation
- Temporal Ensembling Model
- Data Preprocessing
- Experimental Results
- Future work
- Conclusion
- Question and answers

Introduction

- What is semi-supervised learning
 - Semi-supervised machine learning falls in between the supervised and unsupervised learning
 - In semi-supervised learning, an algorithm learns from a dataset that includes both labeled and unlabeled data, usually mostly unlabeled.

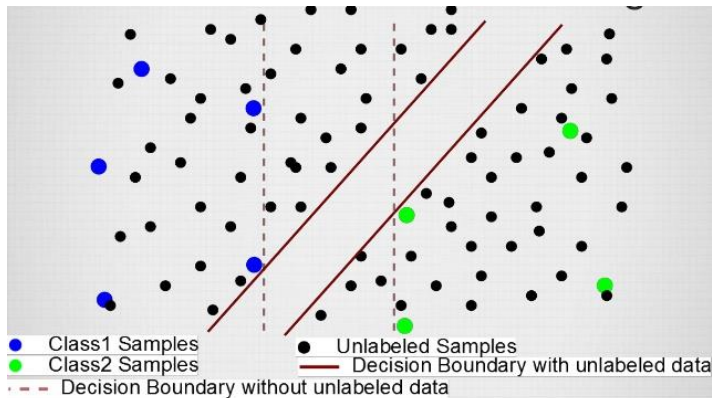


Fig 1: Semi-supervised learning

- Why semi-supervised learning?

Semi-supervised learning takes advantage of a huge amount of unclassified data, to perform a classification in situations when, typically, there is little labeled data which is a representation of where big data is today.



Motivation

- What is Fake news?

It describe factually incorrect and misleading articles published mostly for the purpose of making money through page views.



Fake News Posts Are Mostly Links To Drive Website Traffic

Real vs. Fake News Content Type Distribution on Facebook

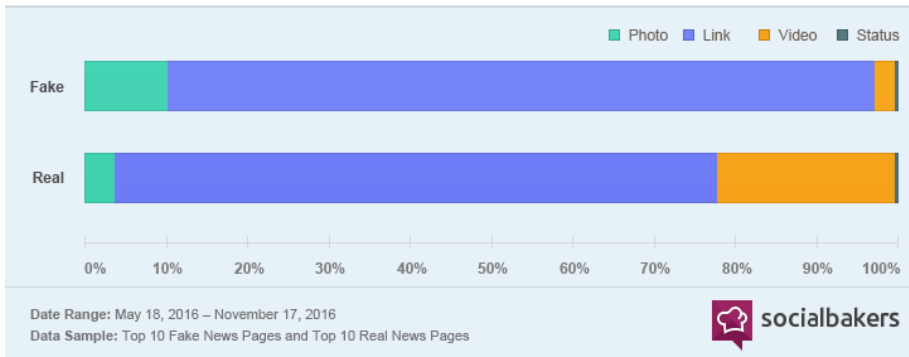


Chart 1: Real vs. Fake news content type distribution on Facebook - <https://www.socialbakers.com>

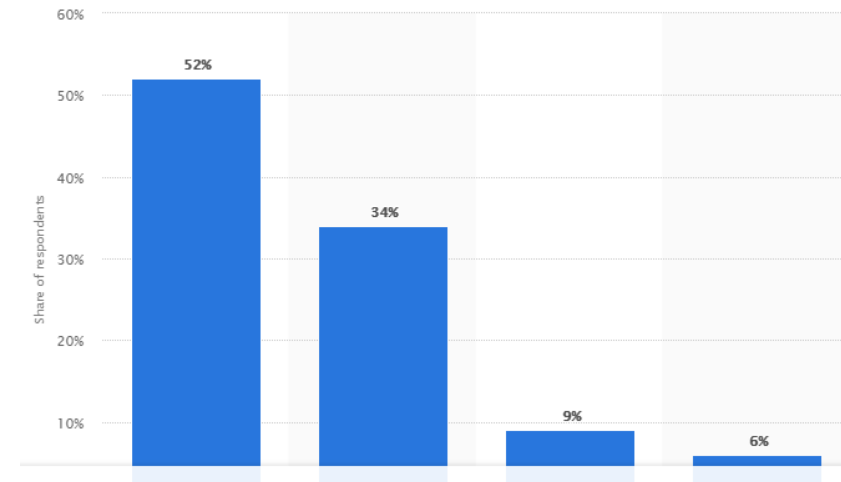


Chart 2: Perceived frequency of online news websites reporting fake news stories in the United States as of March 2018 - <https://www.statista.com>

- Fake news detection

- Understanding on what Fake News is.
- Techniques in the fields of machine learning, natural language processing helps us to detect fake news.



Motivation

- Semi-supervised learning for fake news detection

- *Fake news data is rapidly growing and new words are continuously generated*
- *Insufficient labeled data*
- *It helps to label new data*

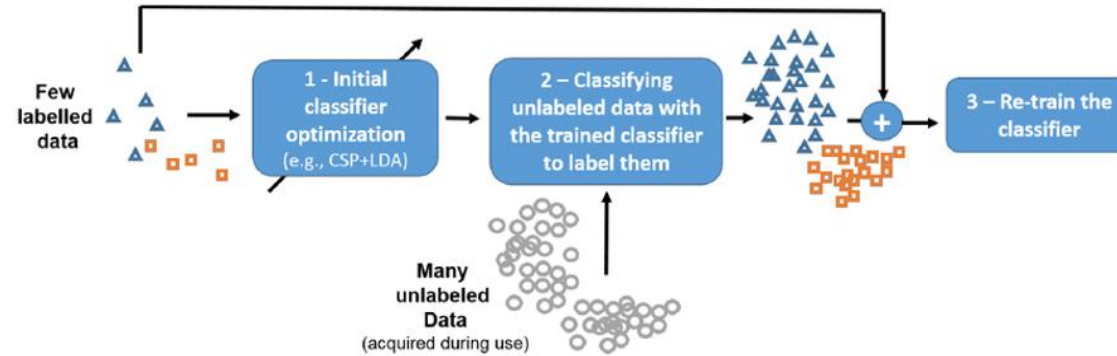


Fig 2: How semi-supervised learning helps to label new data

- Applications

- Government and Politics (Voting and elections)
- Business and Economy (Stock exchange), etc.

Our model for Fake news detection:

Temporal Ensembling model



Temporal Ensembling Model

• Pi Model

- Temporal Ensembling is a variance of the Π -MODEL
- *Encourages consistent network output between two realizations of the same input stimulus, under two different dropout conditions*
 - *Evaluate the network for each training twice*
 - *Loss Function is the sum of two components:*
 - *Component 1. Standard cross entropy loss for labelled*
 - *Component 2. Weighted Mean squared difference of the two prediction vectors (for all inputs)*

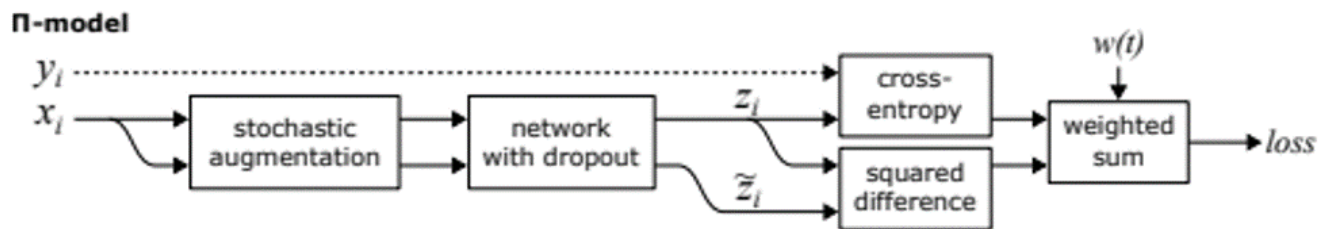


Fig 2: Π -MODEL Diagram Illustration





Temporal Ensembling Model

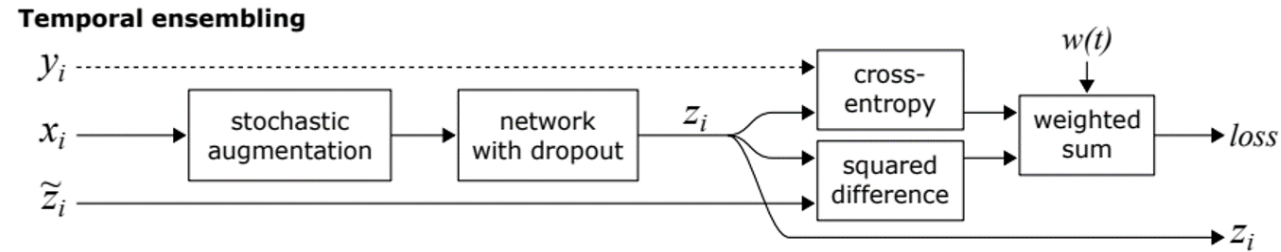


Fig 3: Temporal Ensembling Model Diagram Illustration

- *Advances the model by taking into account the network predictions over multiple previous training epochs*
- *Differs from the Π model in the following ways*
 - Dropout and Augmentations are evaluated just once for every input per epoch
 - and the target vectors for the unsupervised loss component are based on prior network evaluations
 - Faster and less noisy than the Π model





Temporal Ensembling Model

- Evaluate network outputs for augmented inputs
 - Loss Function is the sum of two components:
 - Component 1. Standard cross entropy loss for labelled
 - Component 2. Weighted Mean squared difference of the two prediction vectors (for all inputs)
 - Update Network parameters
 - Construct the target vectors which is then used in training the network on the same inputs for the unsupervised loss component.
 - Accumulate ensemble predictions.



Data Preprocessing



Data Preprocessing Stages

Dataset Information

- We use the sentiment140 dataset. It contains tweets extracted using the twitter API.

Data Set Characteristics:	Tweet messages	Number of Instances:	1,612,839	Area:	Social Media
Attribute Characteristics:	Categorical	Number of Attributes:	varies	Labeled data	12,839
Associated Tasks:	Classification	Missing Values?	nil	Unlabeled data	1,600,000

- Source:** <https://www.kaggle.com/kazanova/sentiment140>



Data Preprocessing Stages

Libraries used

- Some of the libraries used for the data preprocessing includes NumPy, Pandas, matplotlib, Gensim, csv, os, tensorflow, etc

Extracting relevant information(tweets) from data

- ❖ *Pandas was used to extract tweets(unlabeled) from our dataset*
- ❖ *Shell script was used to remove irrelevant information from tweets (links and usernames)*

01467810369	Mon Apr 06 22:19:45 PDT 200	NO_QUERY	TheSpecialOne	@switchfoot http://twitpic.com/2y1zl - Awww, that's a bummer. You shoulda got David Carr of Third Day to do it. ;D	- Awww, that's a bummer. You shoulda got David Carr of Third Day to do it. ;D
01467810672	Mon Apr 06 22:19:49 PDT 200	NO_QUERY	scotthamilton	is upset that he can't update his Facebook by texting it... and might cry as a result School today also. Blah!	is upset that he can't update his Facebook by texting it... and might cry as a result School today also. Blah!
01467810917	Mon Apr 06 22:19:53 PDT 200	NO_QUERY	mattycus	@Kenichan I dived many times for the ball. Managed to save 50% The rest go out of bounds	I dived many times for the ball. Managed to save 50% The rest go out of bounds
01467811184	Mon Apr 06 22:19:57 PDT 200	NO_QUERY	ElleCTF	my whole body feels itchy and like its on fire	my whole body feels itchy and like its on fire
01467811193	Mon Apr 06 22:19:57 PDT 200	NO_QUERY	Karoll	@nationwideclass no, it's not behaving at all. i'm mad. why am i here? because I can't see you all over there.	no, it's not behaving at all. i'm mad. why am i here? because I can't see you all over there.
01467811372	Mon Apr 06 22:20:00 PDT 200	NO_QUERY	joy_wolf	@Kwesidej not the whole crew	not the whole crew
01467811592	Mon Apr 06 22:20:03 PDT 200	NO_QUERY	mybitch	Need a hug	Need a hug
01467811594	Mon Apr 06 22:20:03 PDT 200	NO_QUERY	coZZ	@LOLTrish hey long time no see! Yes.. Rains a bit ,only a bit LOL , I'm fine thanks , how's you ?	hey long time no see! Yes.. Rains a bit ,only a bit LOL , I'm fine thanks , how's you ?
01467811795	Mon Apr 06 22:20:05 PDT 200	NO_QUERY	2Hood4Hollywood	@Tatiana_K nope they didn't have it	nope they didn't have it
01467812025	Mon Apr 06 22:20:09 PDT 200	NO_QUERY	minismo	@twittera que me muera ?	que me muera ?
01467812416	Mon Apr 06 22:20:16 PDT 200	NO_QUERY	erinx3leannexo	spring break in plain city... it's snowing	spring break in plain city... it's snowing
01467812579	Mon Apr 06 22:20:17 PDT 200	NO_QUERY	pardonlauren	I just re-pierced my ears	I just re-pierced my ears
01467812723	Mon Apr 06 22:20:19 PDT 200	NO_QUERY	TLeC	@caregiving I couldn't bear to watch it. And I thought the UA loss was embarrassing	I couldn't bear to watch it. And I thought the UA loss was embarrassing
01467812771	Mon Apr 06 22:20:19 PDT 200	NO_QUERY	robobbierobert	@octolinz16 It it counts, idk why I did either. you never talk to me anymore	
01467812784	Mon Apr 06 22:20:20 PDT 200	NO_QUERY	bayofwolves	@smarrison i would've been the first, but i didn't have a gun. not really though, zac snyder's just a doucheclown.	
01467812799	Mon Apr 06 22:20:20 PDT 200	NO_QUERY	HairByJess	@iamjazzytizzle I wish I got to watch it with you!! I miss you and @iamlinicki how was the premiere?!	
01467812964	Mon Apr 06 22:20:22 PDT 200	NO_QUERY	lovesongwriter	Hollis' death scene will hurt me severely to watch on film why is directors cut not out now?	
01467813137	Mon Apr 06 22:20:25 PDT 200	NO_QUERY	amotley	about to file taxes	
01467813579	Mon Apr 06 22:20:31 PDT 200	NO_QUERY	starkissed	@LettyA ahh ive always wanted to see rent love the soundtrack!!	



Data Preprocessing Stages

Covert categorical labels to numeric

- Used NumPy to convert the target features from categorical to numeric values

statement	label
Building a wall on the U.S.-Mexico border will take literally years.	1
Wisconsin is on pace to double the number of layoffs this year.	0
Says John McCain has done nothing to help the vets.	0
Suzanne Bonamici supports a plan that will cut choice for Medicare Advantage seniors.	2
When asked by a reporter whether hes at the center of a criminal scheme to violate campaign laws, Gov. Scott Walker nodded yes.	4
Over the past five years the federal government has paid out \$601 million in retirement and disability benefits to deceased former federal employees.	1
Says that Tennessee law requires that schools receive half of proceeds – \$31 million per year – from a half-cent increase in the Shelby County sales tax.	1
Says Vice President Joe Biden "admits that the American people are being scammed" with the economic stimulus package.	3
Donald Trump is against marriage equality. He wants to go back.	1
We know that more than half of Hillary Clintons meetings while she was secretary of state were given to major contributors to the Clinton Foundation.	3
We know there are more Democrats in Georgia than Republicans. We know that for a fact.	3
PolitiFact Texas says Congressman Edwards attacks on Bill Flores are false.	3
Denali is the Kenyan word for black power.	4
Says 57 percent of federal spending goes to the military and just 1 percent goes to food and agriculture, including food stamps.	0
On residency requirements for public workers	2
Says the unemployment rate for college graduates is 4.4 percent and over 10 percent for noncollege-educated.	1
Unfortunately we have documented instances where people defecated in the (Statehouse) building.	4
A recent Gallup poll found that 72 percent of Americans and 56 percent of Democrats say the biggest threat to our nations security is big government.	2
"Each year 18 000 people die in America because they don't have health care."	1

❖ {"false": 0, "true": 1, "half-true": 2, "barely-true": 3, "pants-fire": 4, "mostly-true": 5}

Convert corpus (tweets) to vectors

- Using the 'word2vec' from the Gensim model, we convert the corpus(tweets) to vectors using Gensim
- Specific dimensions (100) and window size (5)

```
the 0.069541715 -0.088945076 -0.45617092 0.28801036 -0.098106965 0.05684791 -0.072424375 -0.10608731 0.25199145 -0.23106149 -0.19417709 -0.54537034 -0.21457154 -0.13477
of -0.4200935 -0.10499043 -0.18349967 -0.041444078 -0.03962356 -0.2514173 0.20336474 -0.085575506 0.50084805 -0.09608549 -0.50111777 -0.5130622 0.2557842 -0.16841255 0.
in -0.36231524 -0.16170593 -0.43789688 -0.020071058 -0.0008023831 -0.22973874 0.072208755 -0.20296475 0.14656478 -0.27466205 -0.3729576 -0.4926206 -0.04002383 0.1383991
to -0.1854042 -0.09174699 -0.15046574 0.013760746 -0.16551308 -0.050232933 0.26502395 -0.43506712 -0.17471816 -0.26373053 -0.5216529 -0.7079504 0.13126954 0.42512092 0.
a -0.17411062 0.23345502 -0.13959093 0.22381547 -0.14247976 0.17027883 -0.08129493 -0.33381256 -0.20179476 -0.2076439 -0.11192805 -0.16638418 -0.09372964 0.4223709 0.06
and -0.28172794 0.10260874 -0.135058 -0.09767919 0.019520516 -0.09864881 0.10273859 -0.19951528 -0.10979185 -0.27095747 -0.37425676 -0.4196247 -0.094296165 0.24636957 0
for 0.4143056 -0.02752815 -0.33035383 -0.022847086 -0.1298408 -0.038234748 0.11049696 -0.26826784 0.005616314 -0.3440353 -0.53922004 -0.5773987 0.051371727 0.35509148
that -0.30623913 0.14080411 -0.21386455 0.084667094 0.12181833 -0.0201065 0.14620307 -0.439166 0.020974025 -0.22427416 -0.33000532 -0.34445077 0.09668593 0.3893974 -0.0
is -0.5015464 0.25270572 -0.32904118 -0.05038176 0.1410103 -0.079670176 -0.250891 -0.029440207 0.10308102 -0.3212114 -0.25790513 -0.21762386 0.16593727 -0.095704116 0.0
says -0.43070596 0.026201707 -0.31489512 0.017288223 0.20622309 0.09903601 0.13590664 -0.09761244 -0.031645633 -0.14601041 -0.4968861 -0.072645806 -0.010787154 0.427453
as -0.10063003 0.04744111 -0.17060775 -0.07020707 0.07066000 -0.10160473 0.10704704 -0.16741304 -0.07060000 -0.3130605 -0.44640705 -0.54710114 0.10770300 0.1777600
```



Data Preprocessing Stages

Mapped the vector to matrix

- We represent each tweet by a matrix whose dimensions are "words" by "vectors" using NumPy

```
the 0.069541715 -0.088945076 -0.45617092 0.28801036 -0.098106965 0.05684791 -0.072424375 -0.10608731 0.25199145 -0.23106149 -0.19417709 -0.54537034 -0.21457154 -0.13477
of -0.4200935 -0.10499043 -0.18349987 -0.041444078 -0.03962356 -0.2514173 0.20336474 -0.085575506 0.50084805 -0.09608549 -0.50111777 -0.5130622 0.2557842 -0.16841255 0.
in -0.36231524 -0.16170593 -0.43789688 -0.020071058 -0.0008023831 -0.22973874 0.072208755 -0.20296475 0.14656478 -0.27466205 -0.3729576 -0.4926206 -0.04002383 0.1383991
to -0.1854042 -0.09174699 -0.15046574 0.013760746 -0.16551308 -0.050232933 0.26502395 -0.43506712 -0.17471816 -0.26373053 -0.5216529 -0.7079504 0.13126954 0.42512092 0.
a -0.17411062 0.23345502 -0.13959093 0.22381547 -0.14247976 0.17027883 -0.08129493 -0.33381256 -0.20179476 -0.2076439 -0.11192805 -0.16638418 -0.09372964 0.4223709 0.06
and -0.28172794 0.10260874 -0.135058 -0.09767919 0.019520516 -0.09064881 0.10273859 -0.19951528 -0.10979185 -0.27095747 -0.37425676 -0.4196247 -0.094296165 0.24636957 0
for -0.4143056 -0.02752815 -0.33035383 -0.022847086 -0.1298408 -0.038234748 0.11049696 -0.26826784 0.005616314 -0.3440353 -0.53922004 -0.5773987 0.051371727 0.35509148
that -0.30623913 0.14080411 -0.21386455 0.084667094 0.12181833 -0.0201065 0.14620307 -0.439166 0.020974025 -0.22427416 -0.33000532 -0.34445077 0.09668593 0.3893974 -0.0
is -0.5015464 0.25270572 -0.32904118 -0.05038176 0.1410103 -0.079670176 -0.250891 -0.029440207 0.10308102 -0.3212114 -0.25790513 -0.21762386 0.16593727 -0.095704116 0.0
says -0.43070596 0.026201707 -0.31489512 0.017288223 0.20622309 0.09903601 0.13590664 -0.09761244 -0.031645633 -0.14601041 -0.4968861 -0.072645806 -0.010787154 0.427453
an -0.30063003 0.06754411 -0.17060375 -0.033034303 0.00666000 -0.101604335 0.10704704 -0.16341304 -0.07040000 -0.3130605 -0.44640704 -0.64710114 0.10770300 0.177760 0
```

Covert the data to 'tfrecord' format

- ❖ We finally convert the data to tensor flow data format for the neural network training



Experimental Results

SVHN Dataset

- Still at the final stages of preprocessing of the twitter dataset
- For purposes of the experiment results, we use the SVHN dataset to train and test the model
- The street view house numbers (SVHN) dataset consists of 32×32 pixel RGB images of real-world house numbers
- Task is to classify the centermost digit.

Dataset	Domain	Dimensionality	Training set	Test set
SVHN	Vision	3072(32 x 32)	73k	26k

Results after 300 Epochs

- ❖ Training Accuracy – 72.33% Training Duration: 10 hrs
- ❖ Validation Accuracy – 0.876
- ❖ Final Test Accuracy – 85.85%

```
Epoch 289/300: Train Loss: 1.7595108, Train Accuracy: 73.384473%, Validation Loss: 1.6050423, Validation Accuracy: 85.619048%, lr=0.000000146, unsupervised weight=1.246, beta=0.500292718
Epoch 290/300: Train Loss: 1.7541108, Train Accuracy: 74.603175%, Validation Loss: 1.6026906, Validation Accuracy: 86.095238%, lr=0.000000100, unsupervised weight=1.246, beta=0.500199199
Epoch 291/300: Train Loss: 1.7551061, Train Accuracy: 74.404762%, Validation Loss: 1.5957895, Validation Accuracy: 86.476190%, lr=0.000000067, unsupervised weight=1.246, beta=0.500134170
Epoch 292/300: Train Loss: 1.7474502, Train Accuracy: 73.160173%, Validation Loss: 1.6035540, Validation Accuracy: 85.714280%, lr=0.000000045, unsupervised weight=1.246, beta=0.500089526
Epoch 293/300: Train Loss: 1.7545340, Train Accuracy: 74.206349%, Validation Loss: 1.6102346, Validation Accuracy: 85.230095%, lr=0.000000030, unsupervised weight=1.246, beta=0.500059128
Epoch 294/300: Train Loss: 1.7667983, Train Accuracy: 72.961760%, Validation Loss: 1.5942510, Validation Accuracy: 86.857143%, lr=0.000000019, unsupervised weight=1.246, beta=0.500038624
Epoch 295/300: Train Loss: 1.7555910, Train Accuracy: 74.080087%, Validation Loss: 1.5941752, Validation Accuracy: 86.952381%, lr=0.000000013, unsupervised weight=1.246, beta=0.500025034
Epoch 296/300: Train Loss: 1.7663376, Train Accuracy: 72.871573%, Validation Loss: 1.6057867, Validation Accuracy: 85.523810%, lr=0.000000008, unsupervised weight=1.246, beta=0.500016034
Epoch 297/300: Train Loss: 1.7482791, Train Accuracy: 73.719336%, Validation Loss: 1.5897080, Validation Accuracy: 87.047619%, lr=0.000000005, unsupervised weight=1.246, beta=0.500010192
Epoch 298/300: Train Loss: 1.7321976, Train Accuracy: 74.350649%, Validation Loss: 1.6035435, Validation Accuracy: 86.380952%, lr=0.000000003, unsupervised weight=1.246, beta=0.500006378
Epoch 299/300: Train Loss: 1.7682675, Train Accuracy: 72.799423%, Validation Loss: 1.5931558, Validation Accuracy: 86.857143%, lr=0.000000002, unsupervised weight=1.246, beta=0.500003904
Epoch 300/300: Train Loss: 1.7789709, Train Accuracy: 72.338447%, Validation Loss: 1.5967487, Validation Accuracy: 86.666667%, lr=0.000000001, unsupervised weight=1.246, beta=0.500002444
Train Ended! Best Validation accuracy = 0.8761904761904762
Final Test Accuracy: 85.850575%
```

Method	Epoch	Train. Acc.	Valid Acc.	Test. Acc.
Temporal Ensembling	300	72.33%	87.6%	85.85%



Future work

- Further work on this project would be to test the Temporal Ensembling model on the preprocessed twitter dataset, evaluate the performance and possible carry out research to improve the model performance in fake news detection.
- Research possible methods or ways to improve the time taken in training the model.





Conclusion

A major part of this project was dedicated to data preprocessing as the data had to be converted to a format that our model can communicate with. The next step will be to test our model using our new dataset and seek further ways to improve the model



THANK YOU

Q & A