

# Normalizing tweets with neural language models



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#### **ABSTRACT**

Due to size limitations and genre conventions tweets often contain a large proportion of abbreviations, alternative spellings, novel words and other non-canonical language. These features are problematic for standard language analysis tools, so it can be desirable to normalize tweets, i.e. convert them to canonical form.

- Align original and normalized strings at character level
- Find the shortest edit script which transforms original into normalized
- Treat the edit operation at each position in original string as a label
- Use a Conditional Random Field as a model to learn such labels.
- Use character ngrams as features
- Use learned text embeddings as additional features

The text embeddings are generated using an Simple Recurrent Network (aka Elman Net) as a language model. The embedding at a certain position in a string is the activation of the hidden layer as the network is predicting the character at this position.

The neural language model was train on a raw sample of tweets (414 million bytes), without language-based or any other type of filtering.

We find that enriching the feature set with learned text embeddings substantially lowers word error rates on tweet normalization on two datasets in two languages.

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# NORMALIZATION EXAMPLES

#### I will c wat i can do

i will see what i can do

## imma jus start puttn it out there

imma just start putting it out there

# Buuenoo poos mee voii muu contentaa

Bueno pues me voy muy contenta

#### Hechoo de menos a mi mami :'(

Echo de menos a mi mamá:'(

### **ALIGNMENT**

Original

| Original | Defectors | THISELCIONS | парет        |
|----------|-----------|-------------|--------------|
| W        |           |             | NULL         |
| i        |           |             | NULL         |
| 1        |           |             | NULL         |
| 1        |           |             | NULL         |
|          |           |             | NULL         |
| С        | С         |             | DELETE       |
|          |           | see         | INSERT (see) |
| W        |           |             | NULL         |
| a        |           | h           | INSERT (h)   |
| t        |           |             | NULL         |
|          |           |             |              |

Deletions Insertions Label

#### **N-GRAM FEATURES**

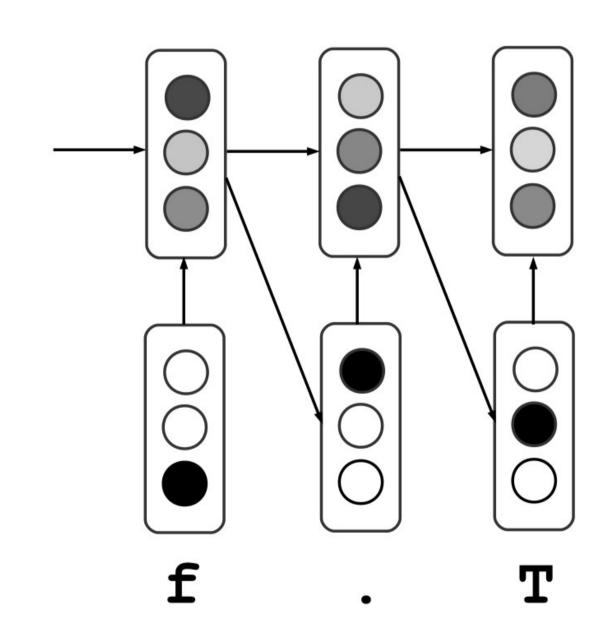
| Position |   | Unigrams | Bigrams | Trigrams |
|----------|---|----------|---------|----------|
| -1       | W |          |         |          |
| 0        | a | w a t    | wa at   | wat      |
| +1       | t |          |         |          |

#### RANDOMLY GENERATED TWEETS

@YuszLAL100A 暇すぎるwwwwwとか麺役者についてる…(ゝ > 晒せ 信じに行けていいんだな…。 RT @yaepdrrafa: @fsch\_chany siaaa,, dobek taha subus sama kiri kaburwanak... hahah なかなかない。 やばい

But I'm the good first-Good Chulc

**ELMAN NET EMBEDDINGS** 



Example nearest neighbors in embedding space

| should h | should d | will s  | will m | should a |
|----------|----------|---------|--------|----------|
| @justth  | @neenu   | @raven_ | @lanae | @despic  |
| maybe u  | maybe y  | cause i | wen i  | when i   |

# 

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