### 5.1 LANGUAGE MODEL

The goal of a language model is to assign a probability to a sequence of words by means of a probability distribution. Formal grammars (e.g. regular, context free) give a hard
-binaryll model of the legal sentences in a language. NLP is a probabilistic model of a language that gives a probability that a string is a member of a language or not. To specify a correct probability distribution, the probability of all sentences in a language must sum to 1 .

### 5.1.1 Uses of Language Models

- Speech Recognition
- OCR \& Handwriting Recognition
- Machine Translation
- Generation
- Context sensitive spelling correction.

A language model also supports predicting the completion of a sentence.
Predictivetext input systems can guess what is been typed and provide choices on how to complete it.

### 5.1.2 N- Gram Word Models

- This model is considered over sequences of words, characters, syllables or other units.
- Estimate probability of each word given prior context.
- An N-gram model uses only $\mathrm{N}-1$ words of prior context.
$\checkmark$ Unigram: P (phone)
$\checkmark$ Bigram: $\mathrm{P}($ phone $\mid$ cell $)$
$\checkmark$ Trigram: P (phone | your cell)
- The Markov assumption is the presumption that the future behavior of a dynamical system only depends on its recent history. In particular, in a Kth-Order Markov Model, next state only depends on the k most recent states, therefore an N - gram model is a ( $\mathrm{N}-1$ ) - order Markov model.


### 5.1.3 N-gram Character Models

- One of the simplest language models: $\boldsymbol{P}\left(\boldsymbol{c}_{1}{ }^{N}\right)$
- Language identification: given the text determine which language it is written in.
- Build a trigram character model of each candidate language: $\boldsymbol{P}\left(\boldsymbol{c}_{\boldsymbol{i}} \mid \boldsymbol{c}_{\boldsymbol{i - 2 i - 1}}, \boldsymbol{l}\right)$
- Train and Test Corpora
$\checkmark$ A language model must be trained on a large corpus of text to estimate goodparameter values.
$\checkmark$ Model can be evaluated based on its ability to predict a high probability for adisjoint test corpus.
$\checkmark$ The training corpus should be representative of the actual application data.
$\checkmark$ To handle words in the test corpus that did not occur in the training data anexplicit symbol is used.
$\checkmark$ Symbol to represent unknown words (<UNK>)
$\checkmark$ Perplexity - Measure of how well a model -fits\|l the test data.

$$
\text { Perplexity }\left(W_{1}^{N}\right)=\sqrt[N]{\frac{1}{P\left(w_{1} w_{2} \ldots w_{N}\right)}}
$$

$\checkmark$ Smoothing-reassigns probability mass to unseen events.

