# Challenges in N-Gram Language Modeling



• N = 3 (Tri-gram model)

P("Jack built that house")

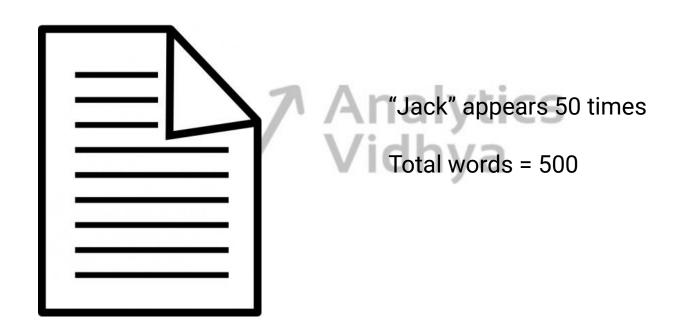




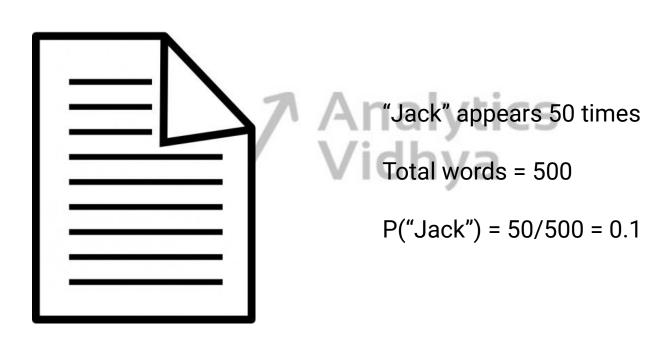
- N = 3 (Tri-gram model)
- P("Jack built that house") =

P("Jack"). P("built"|"Jack"). P("that"|"Jack built"). P("house"|"built that")

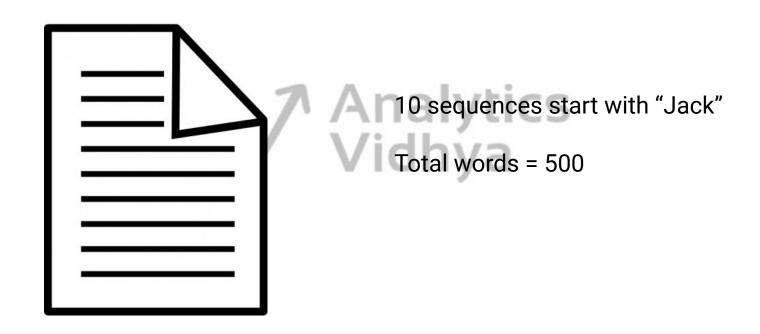
















10 sequences start with "Jack"

Total words = 500

P("Jack") = 10/500 = 0.02



- N = 3 (Tri-gram model)
- P("Jack built that house") =

P("Jack"). P("built"|"Jack"). P("that"|"Jack built"). P("house"|"built that")



- N = 3 (Tri-gram model)
- P("Jack built that house") =

P("Jack"). P("built"|"Jack"). P("that"|"Jack built"). P("house"|"built that")

- Add a start token
  - P(" <s> Jack built that house")



Input = "Jack built that house"





- Input = "Jack built that house"
- Model output = "Jack built that house in two years..."





- Input = "Jack built that house"
- Model output = "Jack built that house in two years in..."





- Input = "Jack built that house"
- Model output = "Jack built that house in two years in India..."





- Input = "Jack built that house"
- Model output = "Jack built that house in two years in India is..."





- Input = "Jack built that house"
- Model output = "Jack built that house in two years in India is..."
- Model does not know when to stop text generation



- N = 3 (Tri-gram model)
- P("Jack built that house") =

P("Jack"). P("built"|"Jack"). P("that"|"Jack built"). P("house"|"built that")

- Add Start Token
  - P(" <s> Jack built that house")
- Add End Token
  - P(" <s> Jack built that house </s>")



- N = 3 (Tri-gram model)
- P("Jack built that house") =

P("Jack"). P("built"|"Jack"). P("that"|"Jack built"). P("house"|"built that")

- Add Start Token
  - P(" <s> Jack built that house")
- Add End Token
  - P(" <s> Jack built that house </s>")

P("Jack"). P("built"|"Jack"). P("that"|"Jack built"). P("house"|"built that") P("</s>"|"that house")



#### Limitations of N-Gram LM

 The higher the N, the better is the model usually. But this leads to lots of computation overhead that requires large computation power.

Vidhya



#### Limitations of N-Gram LM

- The higher the N, the better is the model usually. But this leads to lots of computation overhead that requires large computation power.
- **Sparsity Problem**: N-grams are a sparse representation of language. It will give zero probability to all the words that are not present in the training corpus.



- Training Set
  - ...find the differences
  - ...find the station
  - ...find the jacket





- Training Set
  - ...find the differences
  - ...find the station
  - ...find the jacket
- Unseen Set
  - ...find the cat





- Training Set
  - ...find the differences
  - ...find the station
  - ...find the jacket
- Unseen Set
  - ...find the cat
- P("cat" | "find the") = 0





Laplace Smoothing: Add one to the occurrence of each word





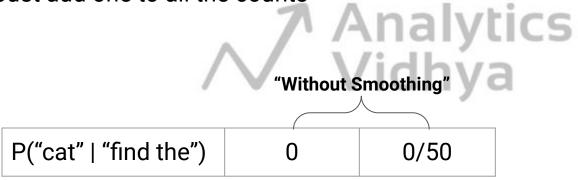
Laplace Smoothing: Add one to the occurrence of each word

Just add one to all the counts



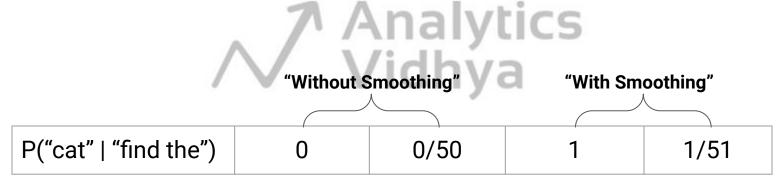


- Laplace Smoothing: Add one to the occurrence of each word
- Just add one to all the counts





- Laplace Smoothing: Add one to the occurrence of each word
- Just add one to all the counts





- Backoff: Switch to shorter context/history as an approximation
  - P("offer"| "denied the") -> P("offer"| "the") -> P("offer")

Analytics Vidhya

- Backoff: Switch to shorter context/history as an approximation
  - P("offer" | "denied the") -> P("offer" | "the") -> P("offer")
- Interpolation: Use weighted sum of unigrams, bigrams, and trigrams

$$P(word_n | word_{n-1} word_{n-2}) = a_1 P(word_n | word_{n-1} word_{n-2})$$

$$+ a_2 P(word_n | word_{n-1})$$

$$+ a_3 P(word_n)$$

$$\circ$$
  $\Sigma a_i = 1$ 

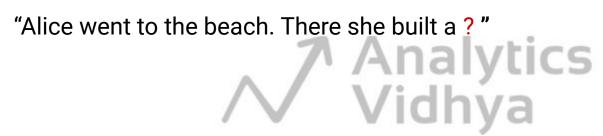


"There she built a?"





- "There she built a?"





- "There she built a?"
- "Alice went to the beach. There she built a sandcastle"







