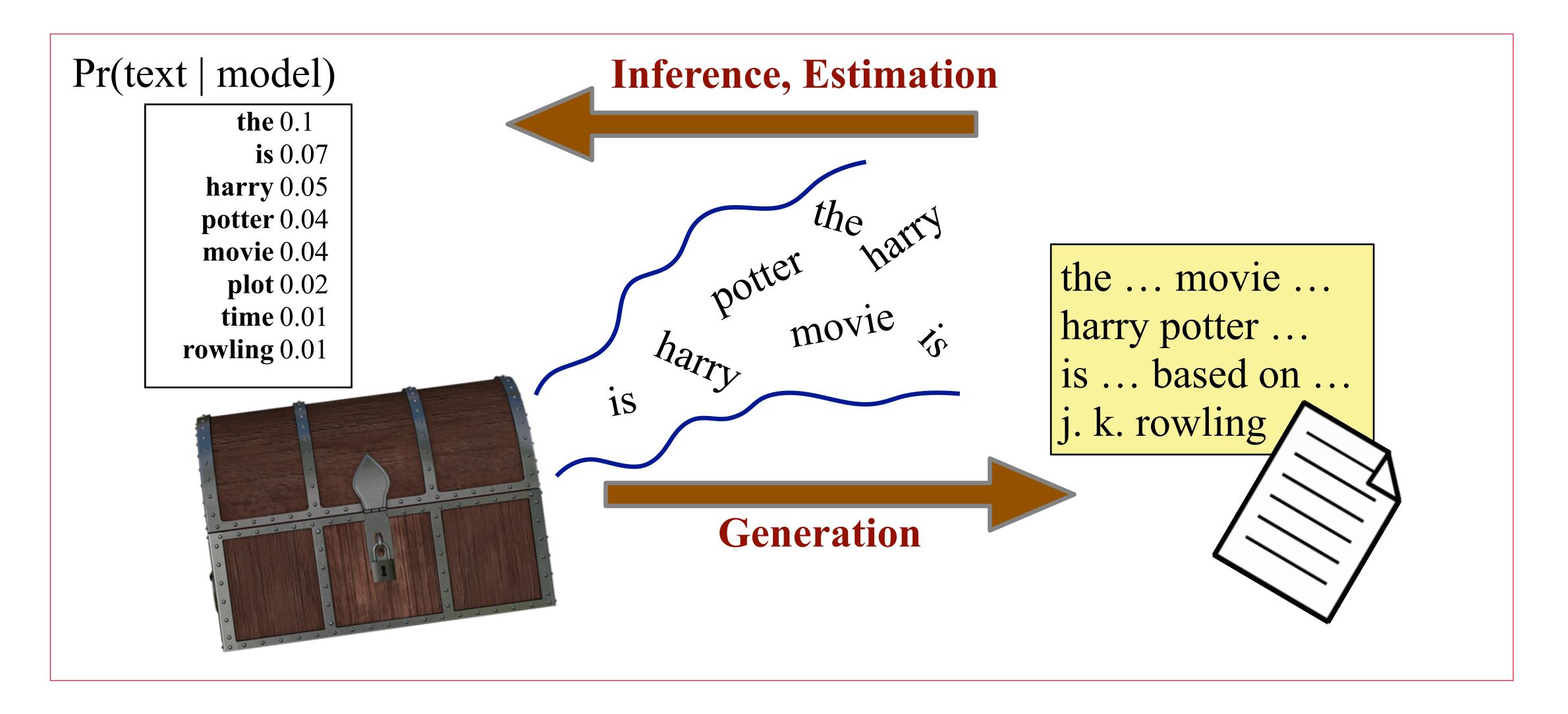


# Applied Text Mining in Python

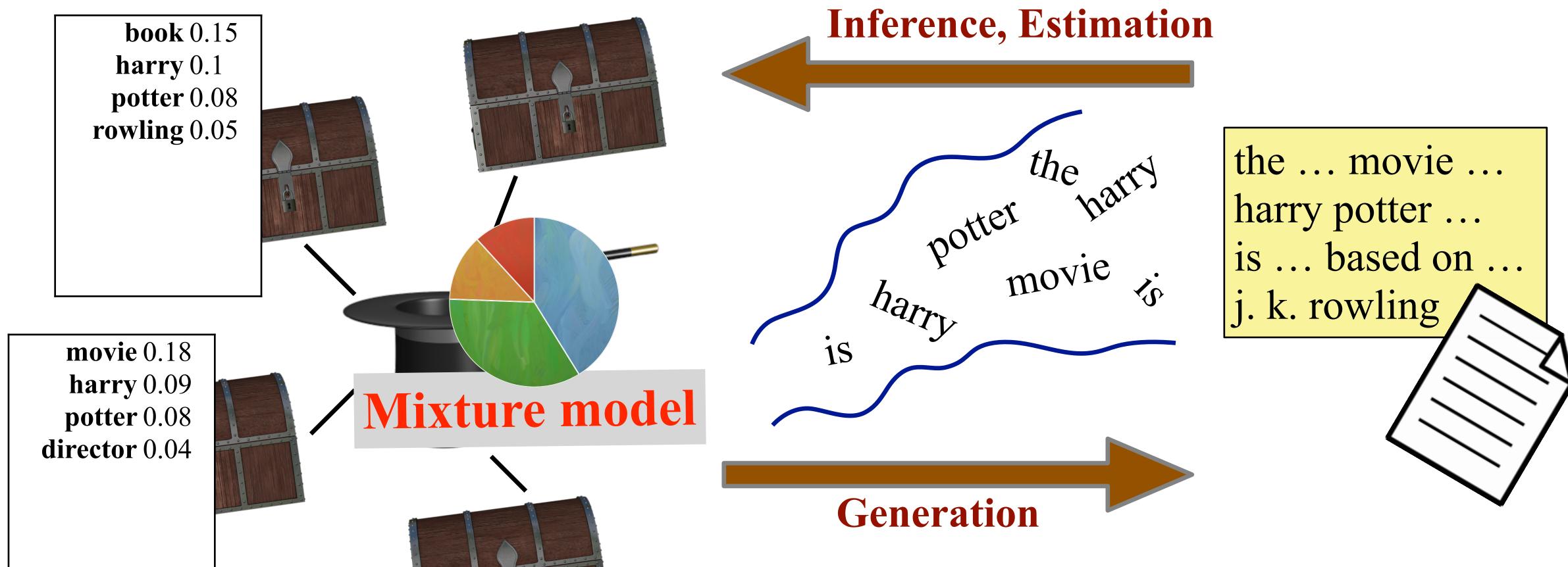
Generative models and LDA

#### Generative Models for Text



#### Generative Models can be Complex

Pr(text | model)





#### Latent Dirichlet Allocation (LDA)

- Generative model for a document d
  - Choose length of document d
  - Choose a mixture of topics for document d
  - Use a topic's multinomial distribution to output words to fill that topic's quota



#### Topic Modeling in Practice

- How many topics?
  - Finding or even guessing the number of topics is hard
- Interpreting topics
  - Topics are just word distributions
  - Making sense of words / generating labels is subjective



### Topic Modeling: Summary

- Great tool for exploratory text analysis
  - What are the documents (tweets, reviews, news articles) about?
- Many tools available to do it effortlessly in Python



#### Working with LDA in Python

- Many packages available, such as gensim, Ida
- Pre-processing text
  - Tokenize, normalize (lowercase)
  - Stop word removal
  - Stemming
- Convert tokenized documents to a document term matrix
- Build LDA models on the doc-term matrix

## Working with LDA in Python (2)

doc\_set: set of pre-processed text documents

```
import gensim
from gensim import corpora, models
dictionary = corpora.Dictionary(doc_set)
corpus = [dictionary.doc2bow(doc) for doc in doc_set]
ldamodel = gensim.models.ldamodel.LdaModel (corpus, num_topics=4, id2word=dictionary, passes=50)
print(ldamodel.print_topics(num_topics=4, num_words=5))
```

Idamodel can also be used to find topic distribution of documents

```
topic_dis = ldamodel[new_doc]
```

#### Take Home Concepts

- Topic modeling is an exploratory tool frequently used for text mining
- Latent Dirichlet Allocation is a generative model used extensively for modeling large text corpora
- LDA can also be used as a feature selection technique for text classification and other tasks