

MOD500 Decision Analysis with Artificial Intelligence Support

Enrico Riccardi¹

Department of Energy Resources, University of Stavanger (UiS).¹

Oct 25, 2024



1 Recap

2 Language models

3 LLMs

Decision trees learning

It is a simple model for supervised classification

Each decision nodes performs a Boolean test (binary split version)

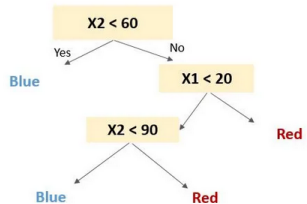
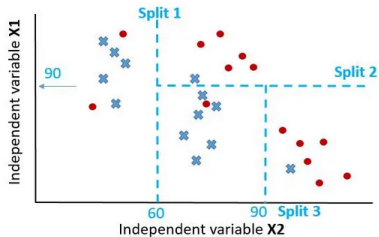
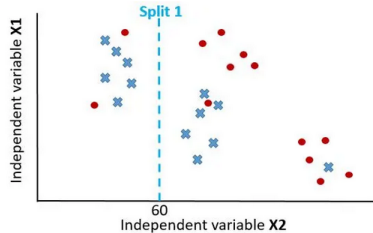
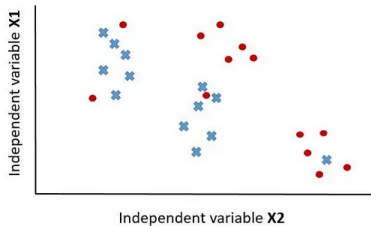
They are build out of DATA!

At each split, we perform the slip that reduce entropy the most.

REMINDER

We need to provide a label!

Decision tree outcome



Pseudo-code

- Compute the entropy of each feature (myopic approach)
- Pick the feature with the maximum entropy
- For each value of the selected feature, compute the entropy of the new population
- Compute the Information Gain by splitting the dataset
- Repeat for the number of desired splits

Decision trees in Python

```
"""  
MOD500 tutorial: Decision tree minimal example  
"""  
import numpy as np  
from matplotlib import pyplot as plt  
  
from sklearn.datasets import load_iris  
from sklearn.tree import DecisionTreeClassifier, plot_tree  
  
iris = load_iris()  
X = iris.data  
y = iris.target  
  
clf = DecisionTreeClassifier(max_leaf_nodes=10,  
                             criterion='entropy')  
clf.fit(X, y)  
  
plot_tree(clf, proportion=True, filled=True)  
plt.show()
```

Generate (at least) 4 different probability distributions

Make a meaningful label, and then make a decision tree from the data generated

(Use the given template to sort out Python programming part if you need)

1 Recap

2 Language models

3 LLMs

Language models

A language model is a probability distribution over sequences of words [1].

Jurafsky and Martin: Speech and Language Processing, 2023

$P(\text{Twinkle twinkle little star, how I wonder what you are.}) = 0.99$
 $P(\text{Twinkle twinkle little moon, how I wonder what you are.}) = 0.75$
 $P(\text{Twinkle twinkle little star, how I what you are.}) = 0.3$ $P(\text{Are you what I wonder I how star, little twinkle, twinkle.}) = 0.02$

Language models

A language model is a probability distribution over sequences of words [1].

Jurafsky and Martin: Speech and Language Processing, 2023

$P(\text{Twinkle twinkle little star, how I wonder what you are.}) = 0.99$
 $P(\text{Twinkle twinkle little moon, how I wonder what you are.}) = 0.75$
 $P(\text{Twinkle twinkle little star, how I what you are.}) = 0.3$ $P(\text{Are you what I wonder I how star, little twinkle, twinkle.}) = 0.02$

Vector representations

Vector representation

- tokenization
- word2vec

	aardvark	...	computer	data	result	pie	sugar	...
cherry	0	...	2	8	9	442	25	...
strawberry	0	...	0	0	1	60	19	...
digital	0	...	1670	1683	85	5	4	...
information	0	...	3325	3982	378	5	13	...

Vector representations

Vector representation

- tokenization
- word2vec

	aardvark	...	computer	data	result	pie	sugar	...
cherry	0	...	2	8	9	442	25	...
strawberry	0	...	0	0	1	60	19	...
digital	0	...	1670	1683	85	5	4	...
information	0	...	3325	3982	378	5	13	...

Vector representations

Vector representation

- tokenization
- word2vec

	aardvark	...	computer	data	result	pie	sugar	...
cherry	0	...	2	8	9	442	25	...
strawberry	0	...	0	0	1	60	19	...
digital	0	...	1670	1683	85	5	4	...
information	0	...	3325	3982	378	5	13	...

Sparse Vector representations

	aardvark	...	computer	data	result	pie	sugar	...
cherry	0	...	2	8	9	442	25	...
strawberry	0	...	0	0	1	60	19	...
digital	0	...	1670	1683	85	5	4	...
information	0	...	3325	3982	378	5	13	...

Table of co-occurrences of the words in Wikipedia

Cherry picking is the act of pointing to individual cases that seem to confirm a particular position while ignoring a significant portion of similar cases or data that may contradict that position.

- One dimension for each word – \rightarrow long
- Many values are 0 – \rightarrow sparse

Sparse Vector representations

	aardvark	...	computer	data	result	pie	sugar	...
cherry	0	...	2	8	9	442	25	...
strawberry	0	...	0	0	1	60	19	...
digital	0	...	1670	1683	85	5	4	...
information	0	...	3325	3982	378	5	13	...

Table of co-occurrences of the words in Wikipedia

Cherry picking is the act of pointing to individual cases that seem to confirm a particular position while ignoring a significant portion of similar cases or data that may contradict that position.

- One dimension for each word – \rightarrow long
- Many values are 0 – \rightarrow sparse

Sparse Vector representations

	aardvark	...	computer	data	result	pie	sugar	...
cherry	0	...	2	8	9	442	25	...
strawberry	0	...	0	0	1	60	19	...
digital	0	...	1670	1683	85	5	4	...
information	0	...	3325	3982	378	5	13	...

Table of co-occurrences of the words in Wikipedia

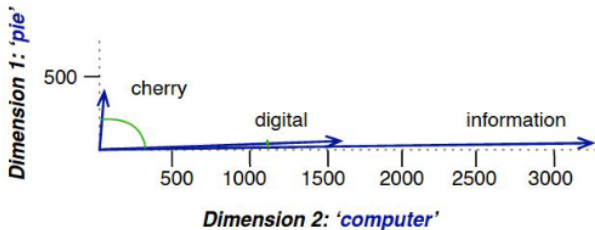
Cherry picking is the act of pointing to individual cases that seem to confirm a particular position while ignoring a significant portion of similar cases or data that may contradict that position.

- One dimension for each word – \rightarrow long
- Many values are 0 – \rightarrow sparse

Vector similarity

Metric alert

How close are two words?



- A neural network designed to explicitly take into account the long-range dependencies between words
- Sequence-to-sequence models that transform an input vectors (x_1, \dots, x_n) to some output vectors (y_1, \dots, y_n) of the same length
- Transformers are made up of stacks of transformer blocks.
- Attention allows to directly extract and use information from arbitrarily long contexts

- A neural network designed to explicitly take into account the long-range dependencies between words
- Sequence-to-sequence models that transform an input vectors (x_1, \dots, x_n) to some output vectors (y_1, \dots, y_n) of the same length
- Transformers are made up of stacks of transformer blocks.
- Attention allows to directly extract and use information from arbitrarily long contexts

Transformers

- A neural network designed to explicitly take into account the long-range dependencies between words
- Sequence-to-sequence models that transform an input vectors (x_1, \dots, x_n) to some output vectors (y_1, \dots, y_n) of the same length
- Transformers are made up of stacks of transformer blocks.
- Attention allows to directly extract and use information from arbitrarily long contexts

- A neural network designed to explicitly take into account the long-range dependencies between words
- Sequence-to-sequence models that transform an input vectors (x_1, \dots, x_n) to some output vectors (y_1, \dots, y_n) of the same length
- Transformers are made up of stacks of transformer blocks.
- Attention allows to directly extract and use information from arbitrarily long contexts

- A neural network designed to explicitly take into account the long-range dependencies between words
- Sequence-to-sequence models that transform an input vectors (x_1, \dots, x_n) to some output vectors (y_1, \dots, y_n) of the same length
- Transformers are made up of stacks of transformer blocks.
- Attention allows to directly extract and use information from arbitrarily long contexts

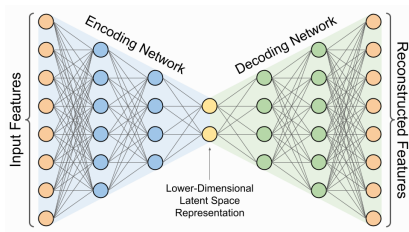
Encode & decode

Encoder model

From an input sequence to a contextualised representation of each input element

Decoder model

From contextualised representations to a task-specific output sequence



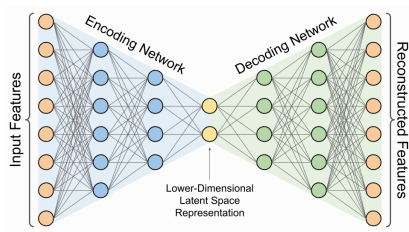
Encode & decode

Encoder model

From an input sequence to a contextualised representation of each input element

Decoder model

From contextualised representations to a task-specific output sequence



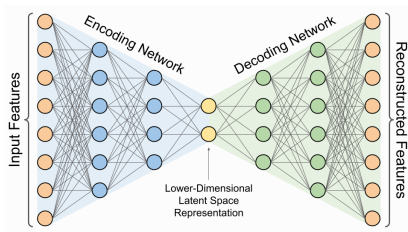
Encode & decode

Encoder model

From an input sequence to a contextualised representation of each input element

Decoder model

From contextualised representations to a task-specific output sequence



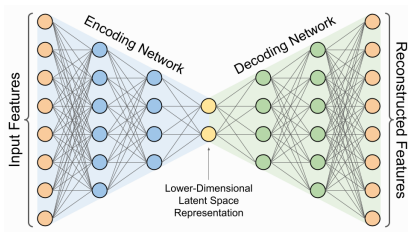
Encode & decode

Encoder model

From an input sequence to a contextualised representation of each input element

Decoder model

From contextualised representations to a task-specific output sequence

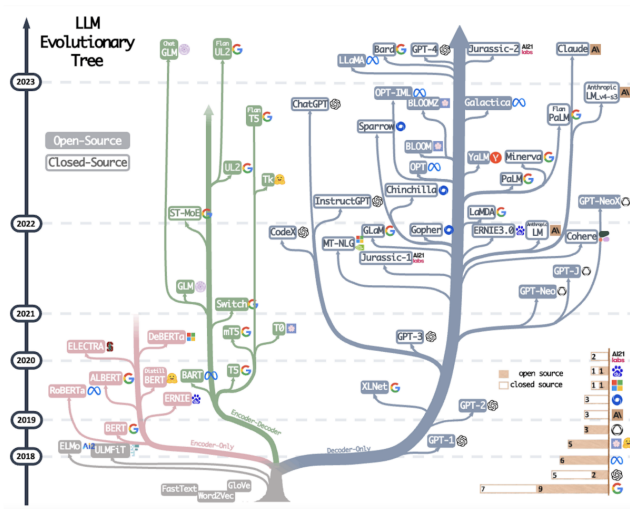


1 Recap

2 Language models

3 LLMs

LLMs



- Speech and Language Processing, Chapter 9 (Transformers) and 10 (Large Language Models), Dan Jurafsky and James H. Martin 17
- The Illustrated Transformer, Jay Alammar