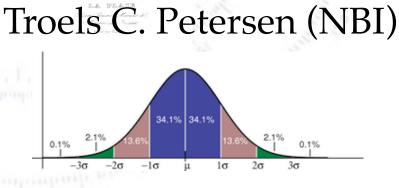
# **Applied ML** Natural Language Processing (NLP)





"Statistics is merely a quantisation of common sense - Machine Learning is a sharpening of it!"

# **Housing Prices**

#### "Inside" vs "outside" variables

Problem with all the variables until now has been that all of the variables are "outside" variables.

(square meter size, distance to X, floor number etc).

We need "inside" variables for extra information

(kitchen condition, bathroom condition etc.)

So we use the descriptions of the houses!

#### **Natural Language Processing**

How to use descriptions:

Flot delevenlig ejerlejlighed på Frederiksbjerg tilbagetrukket fra vejen i hyggeligt baghus!

Køkkenet fremstår pænt og velholdt, og har hårde hvidevarer i stål.

Det sidste værelse i lejligheden er helt sit eget, og har en spændende indretning.

#### **Natural Language Processing**

Multiple documents:

- 1) "The villa is big."
- 2) *"The apartment needs to be renovated."*
- 3) "With a big, newly renovated kitchen."

3,000,000 DKK 1,000,000 DKK 2,000,000 DKK

Remove punctuation Stop words: *the, is, a, to, be* 

### **Natural Language Processing**

- 1) "villa big"
- 2) "apartment needs renovated"
- 3) "with big newly renovated kitchen"

### Natural Language Processing

- 1) "villa big"
- 2) "apartment needs renovated"
- 3) "with big newly renovated kitchen"

Bag of words:

1	apartment	big	kitchen	needs	newly	renovated	villa	with	
1)	0	1	0	0	0	0	1	0	
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### Natural Language Processing

- 1) "villa big"
- 2) "apartment needs renovated"
- 3) "with big newly renovated kitchen"

#### Bag of words:

	apartment	big	kitchen	needs	newly	renovated	villa	with
1)	0	1	0	0	0	0	1	0
2)	1	0	0	1	0	1	0	0
3)	0	1 1	արություն  1	0	1	1	0	1
	18.	180	0	•	•	•	•	

#### **Natural Language Processing**

In [1]: from sklearn.feature\_extraction.text import CountVectorizer

```
corpus = [ 'The villa is big.',
    'The apartment needs to be renovated.',
    'With a big, newly renovated kitchen.']
```

stop\_words = ['the', 'is', 'a', 'to', 'be']

#### **Natural Language Processing**

In [1]: from sklearn.feature\_extraction.text import CountVectorizer

```
corpus = [ 'The villa is big.',
    'The apartment needs to be renovated.',
    'With a big, newly renovated kitchen.']
```

stop\_words = ['the', 'is', 'a', 'to', 'be']

In [2]: vectorizer = CountVectorizer(stop\_words=stop\_words)

bag\_of\_words = vectorizer.fit\_transform(corpus)

```
vocabolary = vectorizer.get_feature_names()
```

```
print(f"\nVocabolary: {vocabolary}")
print(f"Vocabolary lenght: {len(vocabolary)}")
```

```
Vocabolary: ['apartment', 'big', 'kitchen', 'needs', 'newly',
'renovated', 'villa', 'with']
Vocabolary lenght: 8
```

#### **Natural Language Processing**

bag\_of\_words = vectorizer.fit\_transform(corpus)

vocabolary = vectorizer.get\_feature\_names()

```
print(f"\nVocabolary: {vocabolary}")
print(f"Vocabolary lenght: {len(vocabolary)}")
```

Make a (huge but

sparse) matrix of (non-stop) words.

**Natural Language Processing** 

Show the matrix/array.

In [3]: bag\_of\_words.toarray()

Out[3]: array([[0, 1, 0, 0, 0, 0, 1, 0], [1, 0, 0, 1, 0, 1, 0, 0], [0, 1, 1, 0, 1, 1, 0, 1]], dtype=int64)



### **Natural Language Processing**

**Bi-grams**:

vectorizer\_bigram = CountVectorizer(stop\_words=stop\_words( ngram\_range=(1, 2)) )

bag\_of\_words\_bigram = vectorizer\_bigram.fit\_transform(corpus)

vocabolary\_bigram = vectorizer\_bigram.get\_feature\_names()

```
print(f"\nVocabolary: {vocabolary_bigram}")
print(f"Vocabolary lenght: {len(vocabolary_bigram)}")
```

Vocabolary: ['apartment', 'apartment needs', 'big', 'big newly', 'kitchen', 'need s', 'needs renovated', 'newly', 'newly renovated', 'renovated', 'renovated kitche n', 'villa', 'villa big', 'with', 'with big'] Vocabolary lenght: 15

#### **Natural Language Processing**

**Bi-grams**:

bag\_of\_words\_bigram.toarray()

array([[0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0],
 [1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 0],
 [0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 0, 1, 1]], dtype=int64)

#### **Natural Language Processing**

Term Frequency - Inverse Document Frequency: TF-IDF

Natural weighting of words

CountVectorizer, TfidfVectorizer

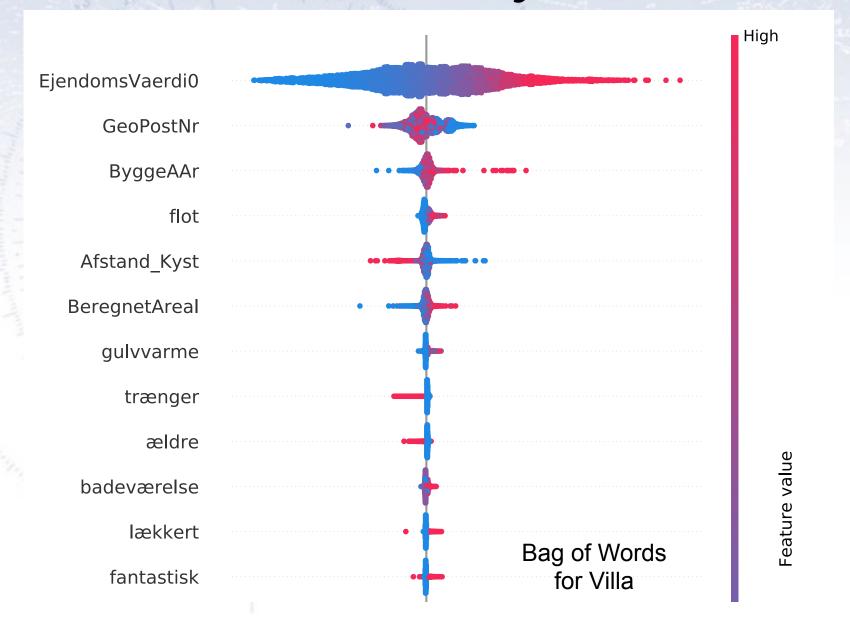
Assign a weight to each word, according to its frequency of use. weight\_IDF = log(N<sub>all</sub> / N<sub>appearances</sub>)

MAD(XGB, numerics only) = 0.165

MAD(XGB, text only, BOW) = 0.254

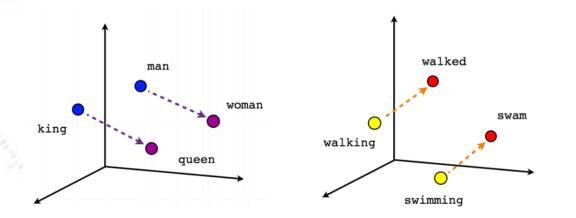
MAD(XGB, combined) = 0.147

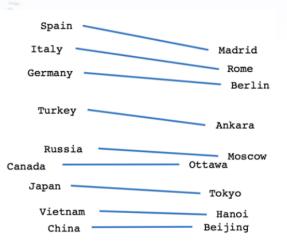
(Numerics: GeoPostNr, BeregnetAreal, ByggeAAr, EjendomsVaerdi0, Afstand\_Kyst)



#### More advanced methods:

- Latent Dirichlet Allocation, LDA
- Word Vectors / Word Embeddings, word2vec, GloVe, FastText





Male-Female

Verb tense

**Country-Capital**