# Extract Transform Select

INTRODUCTION TO SPARK SQL IN PYTHON



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## ETS



# Extract Transform Select

#### Extract, Transform, and Select

- Extraction
- Transformation
- Selection

#### **Built-in functions**

from pyspark.sql.functions import split, explode



#### The length function

```
from pyspark.sql.functions import length
```

```
df.where(length('sentence') == 0)
```



#### Creating a custom function

- User Defined Function
- UDF



#### Importing the udf function

from pyspark.sql.functions import udf



#### Creating a boolean UDF

print(df)

#### DataFrame[textdata: string]

from pyspark.sql.functions import udf

from pyspark.sql.types import BooleanType



#### Creating a boolean UDF

```
short_udf = udf(lambda x:
                          True if not x or len(x) < 10 else False,
                          BooleanType())
df.select(short_udf('textdata')\
  .alias("is short"))\
  .show(3)
|is short|
    false|
    true|
```

false|

#### Important UDF return types

from pyspark.sql.types import StringType, IntegerType, FloatType, ArrayType



#### Creating an array UDF

```
df3.select('word array', in_udf('word array').alias('without endword'))\
    .show(5, truncate=30)
```

```
word array | without endword
[[then, how, many, are, there][then, how, many, are]
              [how, many]|
                                   [how]
          [quite, so]|
                                  [quite]|
  [you, have, not, observed]| [you, have, not]|
```

#### Creating an array UDF

```
from pyspark.sql.types import StringType, ArrayType
```

```
# Removes last item in array
in_udf = udf(lambda x:
    x[0:len(x)-1] if x and len(x) > 1
    else [],
    ArrayType(StringType()))
```

#### Sparse vector format

- 1. Indices
- 2. Values

#### Example:

- Array: [1.0, 0.0, 0.0, 3.0]
- Sparse vector: (4, [0, 3], [1.0, 3.0])

#### Working with vector data

- hasattr(x, "toArray")
- x.numNonzeros())

#### Let's practice!

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# Creating feature data for classification

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#### Transforming a dense array

#### Transforming a dense array

```
try:
    df.select(bad_udf('outvec').alias('label')).first()
except Exception as e:
    print(e.__class__)
    print(e.errmsg)
```

```
<class 'py4j.protocol.Py4JJavaError'>
An error occurred while calling o90.collectToPython.
```

#### UDF return type must be properly cast

#### The UDF in action

```
+----+
|endword| doc|count| features| outvec|
+-----+
| it|[please, do, not,...| 1149|(12847,[15,47,502...| (12847,[7],[1.0])|
| holmes|[start, of, the, ...| 107|(12847,[0,3,183,1...|(12847,[145],[1.0])|
| i|[the, adventures,...| 103|(12847,[0,3,35,14...| (12847,[11],[1.0])|
+-----+
```

```
df.withColumn('label', k_udf('outvec')).drop('outvec').show(3)
```

```
+----+
|endword| doc|count| features|label|
+----+
| it|[please, do, not,...| 1149|(12847,[15,47,502...| 7|
| holmes|[start, of, the, ...| 107|(12847,[0,3,183,1...| 145|
| i|[the, adventures,...| 103|(12847,[0,3,35,14...| 11|
+-----+
```

#### CountVectorizer

- ETS: Extract Transform Select
- CountVectorizer is a Feature Extractor
- Its input is an array of strings
- Its output is a vector

#### Fitting the CountVectorizer



#### Let's practice!

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#### **Text Classification**

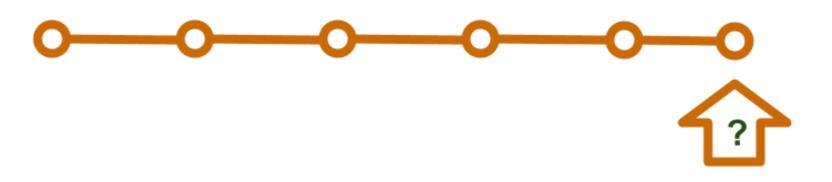
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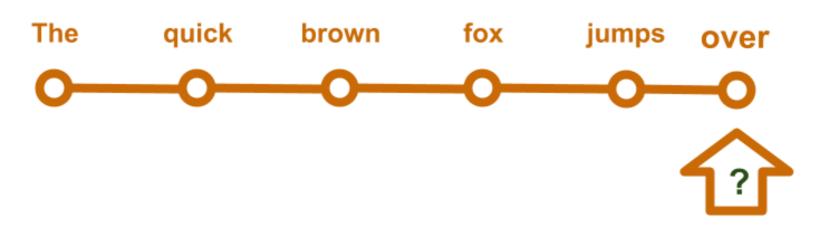


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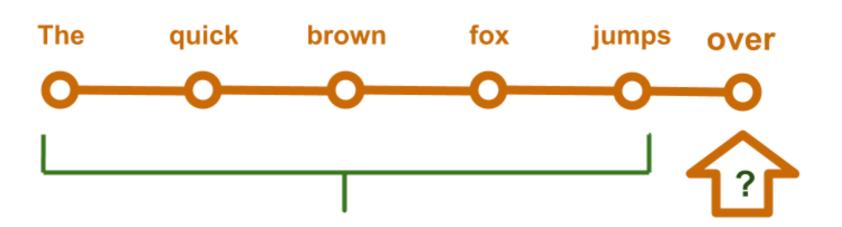


# Endword Prediction

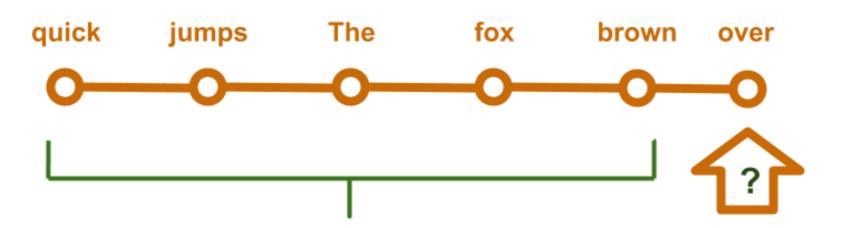




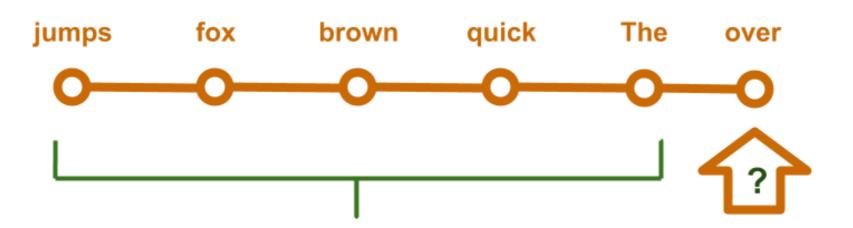




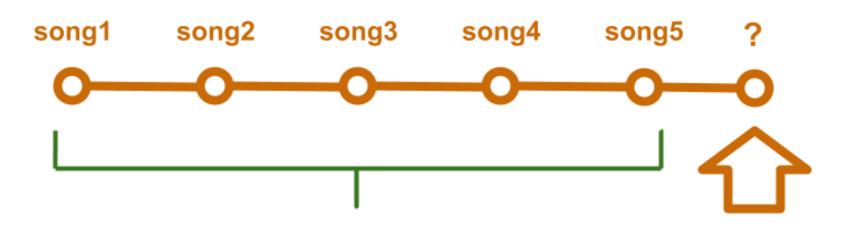




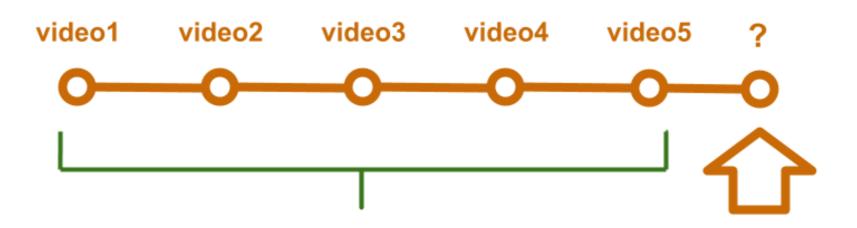












#### Selecting the data

#### Combining the positive and negative data

df\_examples = df\_true.union(df\_false)



#### Splitting the data into training and evaluation sets

df\_train, df\_eval = df\_examples.randomSplit((0.60, 0.40), 42)



#### Training

```
from pyspark.ml.classification import LogisticRegression
logistic = LogisticRegression(maxIter=50, regParam=0.6, elasticNetParam=0.3)
model = logistic.fit(df_train)
print("Training iterations: ", model.summary.totalIterations)
```

#### Let's practice!

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# Predicting and evaluating

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#### Applying a model to evaluation data

```
predicted = df_trained.transform(df_test)
```

- prediction column: double
- probability column: vector of length two

```
x = predicted.first
print("Right!" if x.label == int(x.prediction) else "Wrong")
```

#### Evaluating classification accuracy

```
model_stats = model.evaluate(df_eval)
```

```
type(model_stats)
```

pyspark.ml.classification.BinaryLogisticRegressionSummary)

```
print("\nPerformance: %.2f" % model_stats.areaUnderROC)
```



#### Example of classifying text

- Positive labels:
  - ['her', 'him', 'he', 'she', 'them', 'us', 'they', 'himself', 'herself', 'we']
- Number of examples: 5746
- Number of examples: 2873 positive, 2873 negative
- Number of training examples: 4607
- Number of test examples: 1139
- training iterations: 21
- Test AUC: 0.87

#### Predicting the endword

- Positive label: 'it'
- Number of examples: 438
- Number of examples: 219 positive, 219 negative
- Number of training examples: 340
- Number of test examples: 98
- Test AUC: 0.85

#### Let's practice!

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### Recap INTRODUCTION TO SPARK SQL IN PYTHON



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#### Recap

- Window function SQL
- Extract
- Transform
- Select
- Train
- Predict
- Evaluate

#### Congratulations!

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