06 column-transformer-text-feats

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CPSC 330 Applied Machine Learning

1 Lecture 6: sklearn ColumnTransformer and Text Features

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1.1 Imports

```
[1]: import os
     import sys
     import matplotlib.pyplot as plt
     import numpy as np
     import pandas as pd
     from IPython.display import HTML
     sys.path.append("code/.")
     from plotting_functions import *
     from utils import *
     pd.set_option("display.max_colwidth", 200)
     from sklearn.compose import ColumnTransformer, make_column_transformer
     from sklearn.dummy import DummyClassifier, DummyRegressor
     from sklearn.impute import SimpleImputer
     from sklearn.model_selection import cross_val_score, cross_validate,_
      →train_test_split
     from sklearn.neighbors import KNeighborsClassifier
     from sklearn.pipeline import Pipeline, make_pipeline
     from sklearn.preprocessing import OneHotEncoder, OrdinalEncoder, StandardScaler
     from sklearn.svm import SVC
```

from sklearn.tree import DecisionTreeClassifier

1.2 Learning outcomes

From this lecture, you will be able to

- use ColumnTransformer to build all our transformations together into one object and use it with sklearn pipelines;
- define ColumnTransformer where transformers contain more than one steps;
- explain handle_unknown="ignore" hyperparameter of scikit-learn's OneHotEncoder;
- explain drop="if_binary" argument of OneHotEncoder;
- identify when it's appropriate to apply ordinal encoding vs one-hot encoding;
- explain strategies to deal with categorical variables with too many categories;
- explain why text data needs a different treatment than categorical variables;
- use scikit-learn's CountVectorizer to encode text data;
- explain different hyperparameters of CountVectorizer.

1.3 sklearn's ColumnTransformer

- In most applications, some features are categorical, some are continuous, some are binary, and some are ordinal.
- When we want to develop supervised machine learning pipelines on real-world datasets, very often we want to apply different transformation on different columns.
- Enter sklearn's ColumnTransformer!!
- Let's look at a toy example:

```
[2]: df = pd.read_csv("data/quiz2-grade-toy-col-transformer.csv")
df
```

[2]:	enjoy_course	ml_experience	major	class_attendance	\
0	yes	1	Computer Science	Excellent	
1	yes	1	Mechanical Engineering	Average	
2	yes	0	Mathematics	Poor	
3	no	0	Mathematics	Excellent	
4	yes	0	Psychology	Good	
5	no	1	Economics	Good	
6	yes	1	Computer Science	Excellent	
7	no	0	Mechanical Engineering	Poor	
8	no	0	Linguistics	Average	
9	yes	1	Mathematics	Average	
10	yes	0	Psychology	Good	
11	yes	1	Physics	Average	
12	yes	1	Physics	Excellent	
13	yes	0	Mechanical Engineering	Excellent	
14	no	0	Mathematics	Poor	
15	no	1	Computer Science	Good	

16 17 18 19 20	yes yes no no yes		0 1 1 0 1	C	- P	r Scien Economi Biolo sycholo nguisti	cs gy gy	Average Average Good Poor Excellent
	university_years	lab1	lab2	lab3	lab4	quiz1	quiz	2
0	3	92	93.0	84	91	92	Α Α	
1	2	94	90.0	80	83	91	not A	
2	3	78	85.0	83	80	80	not A	+
3	3	91	NaN	92	91	89	А	
4	4	77	83.0	90	92	85	А	+
5	5	70	73.0	68	74	71	not A	+
6	4	80	88.0	89	88	91	А	+
7	3	95	93.0	69	79	75	not A	+
8	2	97	90.0	94	82	80	not A	+
9	4	95	82.0	94	94	85	not A	+
10	3	98	86.0	95	95	78	Α	+
11	1	95	88.0	93	92	85	A	+
12	2	98	96.0	96	99	100	A	+
13	4	95	94.0	96	95	100	А	+
14	3	95	90.0	93	95	70	not A	+
15	3	92	85.0	67	94	92	not A	+
16	5	75	91.0	93	86	85	Α	+
17	3	86	89.0	65	86	87	not A	+
18	2	91	NaN	90	88	82	not A	+
19	2	77	94.0	87	81	89	not A	+
20	4	96	92.0	92	96	87	Α	+

[3]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 21 entries, 0 to 20
Data columns (total 11 columns):

#	Column	Non-Null Count	Dtype
0	enjoy_course	21 non-null	object
1	ml_experience	21 non-null	int64
2	major	21 non-null	object
3	class_attendance	21 non-null	object
4	university_years	21 non-null	int64
5	lab1	21 non-null	int64
6	lab2	19 non-null	float64
7	lab3	21 non-null	int64
8	lab4	21 non-null	int64
9	quiz1	21 non-null	int64
10	quiz2	21 non-null	object

dtypes: float64(1), int64(6), object(4)

memory usage: 1.9+ KB

1.3.1 Transformations on the toy data

[4]:	df.head()	
------	-----------	--

[4]:	•	enjoy_course	ml_e	xperie	nce			maj	or clas	ss_attendance	\
	0	yes			1	C	omput	er Scien	се	Excellent	
	1	yes			1	Mechani	cal E	Ingineeri	ng	Average	
	2	yes			0		M	[athemati	cs	Poor	
	3	no			0		M	[athemati	cs	Excellent	
	4	yes			0			Psycholo	gу	Good	
		university_ye	ars	lab1	lab2	lab3	lab4	quiz1	quiz2	2	
	0		3	92	93.0	84	91	. 92	A⊣	+	
	1		2	94	90.0	80	83	91	not A-	+	
	2		3	78	85.0	83	80	80	not A-	+	
	3		3	91	NaN	92	91	. 89	A	-	
	4		4	77	83.0	90	92	2 85	Α÷	+	

- Scaling on numeric features
- One-hot encoding on the categorical feature major and binary feature enjoy_course

- Ordinal encoding on the ordinal feature class_attendance
- Imputation on the lab2 feature
- None on the ml_experience feature

1.3.2 ColumnTransformer example

```
Data
```

```
[5]: X = df.drop(columns=["quiz2"])
     y = df["quiz2"]
     X.columns
```

```
[5]: Index(['enjoy_course', 'ml_experience', 'major', 'class_attendance',
            'university_years', 'lab1', 'lab2', 'lab3', 'lab4', 'quiz1'],
           dtype='object')
```

Identify the transformations we want to apply

```
[6]: X.head()
```

[6]:	enjoy_course	ml_experience	major	class_attendance	\
0	yes	1	Computer Science	Excellent	
1	yes	1	Mechanical Engineering	Average	
2	yes	0	Mathematics	Poor	
3	no	0	Mathematics	Excellent	
4	yes	0	Psychology	Good	

university_years lab1 lab2 lab3 lab4 quiz1

```
0
                         92 93.0
                                       84
                                             91
                                                     92
                    2
                                                     91
1
                         94
                             90.0
                                      80
                                             83
2
                   3
                         78
                             85.0
                                      83
                                             80
                                                     80
3
                         91
                               NaN
                                       92
                                             91
                                                     89
4
                         77
                             83.0
                                       90
                                             92
                                                     85
```

For simplicity, let's only focus on scaling and one-hot encoding first.

Create a column transformer

• Each transformation is specified by a name, a transformer object, and the columns this transformer should be applied to.

Convenient make_column_transformer syntax

- Similar to make_pipeline syntax, there is convenient make_column_transformer syntax.
- The syntax automatically names each step based on its class.
- We'll be mostly using this syntax.

[12]: transformed = ct.fit_transform(X)

- When we fit_transform, each transformer is applied to the specified columns and the result of the transformations are concatenated horizontally.
- A big advantage here is that we build all our transformations together into one object, and that way we're sure we do the same operations to all splits of the data.
- Otherwise we might, for example, do the OHE on both train and test but forget to scale the test data.

```
Let's examine the transformed data
```

```
[13]: transformed[:2]
```

```
[13]: array([[-0.09345386,
                              0.3589134 , -0.21733442 ,
                                                          0.36269995,
                                                                        0.84002795,
                0.
                              1.
                                            0.
                                                          0.
                                                                         0.
                                                                     ],
                0.
                              0.
                                            0.
                                                          1.
              [-1.07471942,
                              0.59082668, -0.61420598, -0.85597188, 0.71219761,
                                                          0.
                              0.
                                            0.
                                                                     ]])
                1.
                              0.
                                            0.
```

[14]: type(transformed)

[14]: numpy.ndarray

Note that the returned object is not a dataframe. So there are no column names.

Viewing the transformed data as a dataframe

- How can we view our transformed data as a dataframe?
- We are adding more columns.
- So the original columns won't directly map to the transformed data.
- Let's create column names for the transformed data.

```
[15]: ct.named_transformers_
```

```
[16]: # Here are the new columns created by OneHotEncoder
      ct.named_transformers_["onehotencoder"].get_feature_names_out()
[16]: array(['major_Biology', 'major_Computer Science', 'major_Economics',
             'major_Linguistics', 'major_Mathematics',
             'major_Mechanical Engineering', 'major_Physics',
             'major_Psychology'], dtype=object)
[17]: column_names = (
          numeric feats
          + ct.named_transformers_["onehotencoder"].get_feature_names_out().tolist()
          + passthrough feats
      column names
[17]: ['university_years',
       'lab1',
       'lab3',
       'lab4',
       'quiz1',
       'major Biology',
       'major Computer Science',
       'major_Economics',
       'major_Linguistics',
       'major_Mathematics',
       'major_Mechanical Engineering',
       'major_Physics',
       'major_Psychology',
       'ml_experience']
```

Note that the order of the columns in the transformed data depends upon the order of the features we pass to the ColumnTransformer and can be different than the order of the features in the original dataframe.

```
[18]: pd.DataFrame(transformed, columns=column names).head()
[18]:
        university_years
                               lab1
                                         lab3
                                                   lab4
                                                            quiz1
                                                                  major_Biology \
                -0.093454 0.358913 -0.217334 0.362700 0.840028
      0
                                                                             0.0
      1
                -1.074719 0.590827 -0.614206 -0.855972 0.712198
                                                                             0.0
      2
                -0.093454 -1.264480 -0.316552 -1.312974 -0.693936
                                                                             0.0
      3
                -0.093454 0.242957 0.576409 0.362700 0.456537
                                                                             0.0
                0.887812 -1.380436  0.377973  0.515034 -0.054784
                                                                             0.0
        major_Computer Science major_Economics major_Linguistics \
      0
                            1.0
                                             0.0
                            0.0
                                             0.0
      1
                                                                0.0
                            0.0
                                             0.0
      2
                                                                0.0
      3
                                             0.0
                                                                0.0
                            0.0
```

4		0.0	0.0	0.0	
	major_Mathematics	major_Mechanical	Engineering	major_Physics	\
0	0.0		0.0	0.0	
1	0.0		1.0	0.0	
2	1.0		0.0	0.0	
3	1.0		0.0	0.0	
4	0.0		0.0	0.0	
	major_Psychology	ml_experience			
0	0.0	1.0			
1	0.0	1.0			
2	0.0	0.0			
3	0.0	0.0			
4	1.0	0.0			

ColumnTransformer: Transformed data Adapted from here

Training models with transformed data

• We can now pass the ColumnTransformer object as a step in a pipeline.

```
[19]: # Make a pipeline that applies ct to columns and then SVC to the resulting table
pipe = make_pipeline(ct, SVC())
pipe.fit(X, y)
pipe.predict(X)
```

```
[19]: array(['A+', 'not A+', 'not A+', 'A+', 'A+', 'not A+', 'A+', 'not A+', 'not A+', 'A+', 'A+', 'A+', 'A+', 'not A+', 'not A+', 'not A+', 'not A+', 'not A+', 'A+'], dtype=object)
```

1.3.3 Questions for you

True/False: ColumnTransformer

- 1. You could carry out cross-validation by passing a ColumnTransformer object to cross_validate.
- 2. After applying column transformer, the order of the columns in the transformed data has to be the same as the order of the columns in the original data.
- 3. After applying a column transformer, the transformed data is always going to be of different shape than the original data.
- 4. When you call fit_transform on a ColumnTransformer object, you get a numpy ndarray.

What transformations on what columns? Consider the feature columns below.

• What transformations would you apply on each column?

colour	location	shape	water_content	weight
$\overline{\mathrm{red}}$	canada	NaN	84	100
yellow	mexico	long	75	120
orange	spain	\overline{NaN}	90	NaN
magenta	china	round	NaN	600
purple	austria	NaN	80	115
purple	turkey	oval	78	340
green	mexico	oval	83	NaN
blue	canada	round	73	535
brown	china	NaN	NaN	1743
yellow	mexico	oval	83	265

1.4 More on feature transformations

1.4.1 Multiple transformations in a transformer

• Recall that lab2 has missing values.

```
[20]: X.head()
        enjoy_course ml_experience
[20]:
                                                         major class_attendance \
                                              Computer Science
                                                                       Excellent
      0
                  yes
      1
                                    1
                                       Mechanical Engineering
                  yes
                                                                          Average
      2
                                                   Mathematics
                                    0
                                                                             Poor
                  yes
      3
                                    0
                                                   Mathematics
                                                                       Excellent
                   no
      4
                  yes
                                    0
                                                    Psychology
                                                                             Good
         university_years
                            lab1
                                   lab2
                                         lab3
                                                lab4
                                                      quiz1
      0
                                   93.0
                                           84
                                                  91
                                                          92
                         3
                               92
      1
                         2
                               94
                                   90.0
                                           80
                                                  83
                                                         91
      2
                         3
                               78
                                   85.0
                                           83
                                                  80
                                                         80
      3
                         3
                                           92
                                                  91
                                                         89
                               91
                                    NaN
      4
                         4
                                   83.0
                               77
                                            90
                                                  92
                                                         85
```

- So we would like to apply more than one transformations on it: imputation and scaling.
- We can treat lab2 separately, but we can also include it into numeric_feats and apply both transformations on all numeric columns.

```
[21]: numeric_feats = [
          "university_years",
          "lab1",
          "lab2",
          "lab3",
          "lab4",
          "quiz1",
] # apply scaling
categorical_feats = ["major"] # apply one-hot encoding
```

```
passthrough_feats = ["ml_experience"] # do not apply any transformation
drop_feats = ["class_attendance", "enjoy_course"]
```

• To apply more than one transformations we can define a pipeline inside a column transformer to chain different transformations.

```
[22]: ct = make column transformer(
              make_pipeline(SimpleImputer(), StandardScaler()),
              numeric feats,
          ), # scaling on numeric features
          (OneHotEncoder(), categorical_feats), # OHE on categorical features
          ("passthrough", passthrough_feats), # no transformations on the binary_
       \hookrightarrow features
          ("drop", drop_feats), # drop the drop features
[23]: X_transformed = ct.fit_transform(X)
[24]: column_names = (
          numeric feats
          + ct.named_transformers_["onehotencoder"].get_feature_names_out().tolist()
          + passthrough_feats
      column_names
[24]: ['university_years',
       'lab1',
       'lab2',
       'lab3'.
       'lab4',
       'quiz1',
       'major_Biology',
       'major_Computer Science',
       'major_Economics',
       'major_Linguistics',
       'major_Mathematics',
       'major_Mechanical Engineering',
       'major_Physics',
       'major_Psychology',
       'ml_experience']
[25]: pd.DataFrame(X_transformed, columns=column_names).head()
[25]:
                                           lab2
                                                     lab3
                                                                lab4
                                                                         quiz1 \
         university_years
                                lab1
                -0.093454   0.358913   0.893260   -0.217334   0.362700   0.840028
      0
                -1.074719 \quad 0.590827 \quad 0.294251 \ -0.614206 \ -0.855972 \quad 0.712198
      1
      2
                -0.093454 -1.264480 -0.704099 -0.316552 -1.312974 -0.693936
```

```
3
          -0.093454 0.242957 0.000000 0.576409 0.362700 0.456537
4
           0.887812 - 1.380436 - 1.103439 \ 0.377973 \ 0.515034 - 0.054784
   major_Biology
                  major_Computer Science major_Economics major_Linguistics \
0
             0.0
                                                         0.0
                                                                             0.0
             0.0
                                       0.0
                                                         0.0
                                                                             0.0
1
2
             0.0
                                       0.0
                                                         0.0
                                                                             0.0
             0.0
                                                         0.0
3
                                       0.0
                                                                             0.0
             0.0
4
                                       0.0
                                                         0.0
                                                                             0.0
   major_Mathematics major_Mechanical Engineering major_Physics
0
                  0.0
                                                 0.0
1
                  0.0
                                                 1.0
                                                                 0.0
                                                                 0.0
2
                  1.0
                                                 0.0
3
                  1.0
                                                 0.0
                                                                 0.0
4
                  0.0
                                                 0.0
                                                                 0.0
   major_Psychology ml_experience
0
                 0.0
                                1.0
                 0.0
1
                                1.0
2
                 0.0
                                0.0
3
                 0.0
                                0.0
4
                 1.0
                                0.0
```

1.4.2 sklearn set_config

- With multiple transformations in a column transformer, it can get tricky to keep track of everything happening inside it.
- We can use set_config to display a diagram of this.

```
['class_attendance', 'enjoy_course'])])
```

1.4.3 Incorporating ordinal feature class_attendance

- The class_attendance column is different than the major column in that there is some ordering of the values.
 - Excellent > Good > Average > Poor

```
[29]: X.head()
```

```
[29]:
        enjoy_course
                       ml_experience
                                                          major class_attendance
                                              Computer Science
                                                                        Excellent
      0
                  yes
                                     1
      1
                                       Mechanical Engineering
                                     1
                                                                          Average
                  yes
      2
                                    0
                                                   Mathematics
                  yes
                                                                             Poor
      3
                                    0
                                                   Mathematics
                                                                        Excellent
                   no
      4
                                    0
                                                    Psychology
                                                                              Good
                  yes
         university_years
                                   lab2 lab3
                                               lab4
                                                       quiz1
                            lab1
      0
                         3
                               92
                                   93.0
                                            84
                                                  91
                                                          92
      1
                         2
                               94
                                   90.0
                                            80
                                                  83
                                                          91
      2
                         3
                               78
                                   85.0
                                            83
                                                  80
                                                          80
      3
                         3
                               91
                                    NaN
                                            92
                                                  91
                                                          89
                         4
                                   83.0
                                            90
                                                          85
      4
                               77
                                                  92
```

Let's try applying OrdinalEncoder on class_attendance column.

```
[31]: X_toy.join(X_toy_ord_df).head(10)
[31]:
        class attendance class attendance ord
               Excellent
                                              1.0
      0
      1
                                              0.0
                  Average
      2
                                              3.0
                     Poor
      3
               Excellent
                                              1.0
      4
                     Good
                                              2.0
      5
                     Good
                                              2.0
      6
               Excellent
                                              1.0
      7
                     Poor
                                              3.0
      8
                  Average
                                              0.0
      9
                  Average
                                              0.0
        • What's the problem here?
             - The encoder doesn't know the order.
        • We can examine unique categories manually, order them based on our intuitions, and
          then provide this human knowledge to the transformer.
     What are the unique categories of class attendance?
[32]: X_toy["class_attendance"].unique()
[32]: array(['Excellent', 'Average', 'Poor', 'Good'], dtype=object)
     Let's order them manually.
[33]: class_attendance_levels = ["Poor", "Average", "Good", "Excellent"]
     Note that if you use the reverse order of the categories, it wouldn't matter.
     Let's make sure that we have included all categories in our manual ordering.
[34]: assert set(class_attendance_levels) == set(X_toy["class_attendance"].unique())
[35]: oe = OrdinalEncoder(categories=[class_attendance_levels], dtype=int)
      oe.fit(X_toy[["class_attendance"]])
      ca_ord = oe.transform(X_toy[["class_attendance"]])
      ca_ord_df = pd.DataFrame(
          data=ca_ord, columns=["class_attendance_ord"], index=X_toy.index
      print(oe.categories_)
      X_toy.join(ca_ord_df).head(10)
      [array(['Poor', 'Average', 'Good', 'Excellent'], dtype=object)]
[35]:
        class_attendance_class_attendance_ord
               Excellent
                                                3
      0
                                                1
      1
                  Average
```

```
2
                Poor
                                            0
3
          Excellent
                                            3
4
                Good
                                            2
                                            2
5
                Good
6
          Excellent
                                            3
7
                Poor
                                            0
8
            Average
                                            1
9
            Average
                                             1
```

The encoded categories are looking better now!

More than one ordinal columns?

- We can pass the manually ordered categories when we create an OrdinalEncoder object as a list of lists.
- If you have more than one ordinal columns
 - manually create a list of ordered categories for each column
 - pass a list of lists to OrdinalEncoder, where each inner list corresponds to manually created list of ordered categories for a corresponding ordinal column.

Now let's incorporate ordinal encoding of class_attendance in our column transformer.

```
[38]: ct
[38]: ColumnTransformer(transformers=[('pipeline',
                                        Pipeline(steps=[('simpleimputer',
                                                         SimpleImputer()),
                                                         ('standardscaler',
                                                         StandardScaler())]),
                                        ['university_years', 'lab1', 'lab2', 'lab3',
                                         'lab4', 'quiz1']),
                                       ('onehotencoder', OneHotEncoder(), ['major']),
                                       ('ordinalencoder',
                                        OrdinalEncoder(categories=[['Poor', 'Average',
                                                                     'Good',
                                                                     'Excellent']],
                                                       dtype=<class 'int'>),
                                        ['class_attendance']),
                                       ('passthrough', 'passthrough',
                                        ['ml_experience']),
                                       ('drop', 'drop', ['enjoy_course'])])
[39]: X_transformed = ct.fit_transform(X)
[40]: column names = (
          numeric_feats
          + ct.named_transformers_["onehotencoder"].get_feature_names_out().tolist()
          + ordinal feats
          + passthrough_feats
      column_names
[40]: ['university_years',
       'lab1',
       'lab2',
       'lab3',
       'lab4',
       'quiz1',
       'major_Biology',
       'major_Computer Science',
       'major_Economics',
       'major_Linguistics',
       'major_Mathematics',
       'major_Mechanical Engineering',
       'major_Physics',
       'major_Psychology',
       'class_attendance',
```

16

17

18

0.0

0.0

1.0

```
[41]: pd.DataFrame(X transformed, columns=column names)
[41]:
                                        lab2
                                                           lab4
                                                                    quiz1
         university_years
                              lab1
                                                 lab3
                                                                         \
     0
                -0.093454
                         0.358913
                                   0.893260 -0.217334
                                                      0.362700
                                                                0.840028
     1
                -1.074719 0.590827
                                    0.294251 -0.614206 -0.855972
                                                                 0.712198
     2
                -0.093454 -1.264480 -0.704099 -0.316552 -1.312974 -0.693936
     3
                -0.093454 0.242957
                                   0.000000
                                            0.576409 0.362700
                                                                 0.456537
                 0.887812 -1.380436 -1.103439
     4
                                             0.377973 0.515034 -0.054784
     5
                 1.869077 -2.192133 -3.100139 -1.804821 -2.226978 -1.844409
     6
                 0.887812 -1.032566 -0.105089 0.278755 -0.094302 0.712198
     7
                8
                -1.074719 0.938697
                                   0.294251 0.774844 -1.008306 -0.693936
     9
                 10
                -0.093454
                         1.054653 -0.504429 0.874062 0.972036 -0.949597
     11
                         0.706783 -0.105089
                                             0.675627 0.515034 -0.054784
                -2.055985
     12
                          1.054653
                                             0.973280
                -1.074719
                                    1.492270
                                                      1.581372 1.862671
     13
                 0.887812 0.706783
                                   1.092930
                                             0.973280 0.972036
                                                                1.862671
     14
                -0.093454 0.706783
                                   0.294251
                                             0.675627
                                                       0.972036 -1.972240
     15
                -0.093454 0.358913 -0.704099 -1.904039 0.819702
                                                                0.840028
     16
                 1.869077 -1.612349
                                   0.493921 0.675627 -0.398970 -0.054784
     17
                                    0.094581 -2.102474 -0.398970
                -0.093454 -0.336826
                                                                0.200876
     18
                -1.074719 0.242957
                                    0.000000
                                            0.377973 -0.094302 -0.438275
     19
                -1.074719 -1.380436
                                    1.092930
                                             0.080319 -1.160640
                                                                 0.456537
                 0.887812 0.822740 0.693590
     20
                                            0.576409 1.124370 0.200876
         major_Biology
                       major_Computer Science
                                              major_Economics
                                                              major_Linguistics
     0
                   0.0
                                          1.0
                                                          0.0
                                                                            0.0
     1
                   0.0
                                          0.0
                                                          0.0
                                                                            0.0
     2
                   0.0
                                          0.0
                                                          0.0
                                                                            0.0
     3
                   0.0
                                          0.0
                                                          0.0
                                                                            0.0
                   0.0
     4
                                          0.0
                                                          0.0
                                                                            0.0
     5
                   0.0
                                                                            0.0
                                          0.0
                                                          1.0
     6
                   0.0
                                          1.0
                                                          0.0
                                                                            0.0
     7
                   0.0
                                          0.0
                                                          0.0
                                                                            0.0
     8
                   0.0
                                          0.0
                                                          0.0
                                                                            1.0
     9
                   0.0
                                          0.0
                                                          0.0
                                                                            0.0
     10
                   0.0
                                          0.0
                                                          0.0
                                                                            0.0
     11
                   0.0
                                          0.0
                                                          0.0
                                                                            0.0
     12
                   0.0
                                          0.0
                                                          0.0
                                                                            0.0
     13
                   0.0
                                          0.0
                                                          0.0
                                                                            0.0
     14
                   0.0
                                          0.0
                                                          0.0
                                                                            0.0
     15
                   0.0
                                          1.0
                                                          0.0
                                                                            0.0
```

1.0

0.0

0.0

0.0

1.0

0.0

0.0

0.0

0.0

```
0.0
                                          0.0
                                                             0.0
                                                                                  0.0
19
               0.0
20
                                          0.0
                                                             0.0
                                                                                  1.0
    major_Mathematics
                         major_Mechanical Engineering
                                                          major_Physics \
                                                                      0.0
0
                    0.0
                    0.0
                                                     1.0
                                                                      0.0
1
2
                    1.0
                                                     0.0
                                                                      0.0
3
                    1.0
                                                     0.0
                                                                      0.0
4
                    0.0
                                                     0.0
                                                                      0.0
5
                    0.0
                                                     0.0
                                                                      0.0
6
                    0.0
                                                     0.0
                                                                      0.0
7
                    0.0
                                                     1.0
                                                                      0.0
                                                                      0.0
8
                    0.0
                                                     0.0
                                                     0.0
9
                    1.0
                                                                      0.0
10
                    0.0
                                                     0.0
                                                                      0.0
                                                                      1.0
11
                    0.0
                                                     0.0
12
                    0.0
                                                     0.0
                                                                      1.0
13
                    0.0
                                                     1.0
                                                                      0.0
14
                    1.0
                                                     0.0
                                                                      0.0
15
                    0.0
                                                     0.0
                                                                      0.0
16
                    0.0
                                                     0.0
                                                                      0.0
17
                    0.0
                                                     0.0
                                                                      0.0
18
                    0.0
                                                     0.0
                                                                      0.0
19
                                                                      0.0
                    0.0
                                                     0.0
20
                    0.0
                                                     0.0
                                                                      0.0
    major_Psychology
                        class_attendance ml_experience
                   0.0
0
                                       3.0
                                                        1.0
                   0.0
                                                        1.0
1
                                       1.0
2
                  0.0
                                       0.0
                                                        0.0
3
                   0.0
                                                        0.0
                                       3.0
4
                   1.0
                                       2.0
                                                        0.0
5
                   0.0
                                       2.0
                                                        1.0
6
                   0.0
                                       3.0
                                                        1.0
7
                   0.0
                                       0.0
                                                        0.0
8
                   0.0
                                       1.0
                                                        0.0
9
                   0.0
                                       1.0
                                                        1.0
10
                   1.0
                                                        0.0
                                       2.0
                   0.0
                                                        1.0
11
                                       1.0
12
                   0.0
                                       3.0
                                                        1.0
                   0.0
                                       3.0
                                                        0.0
13
14
                   0.0
                                       0.0
                                                        0.0
15
                  0.0
                                       2.0
                                                        1.0
16
                  0.0
                                                        0.0
                                       1.0
17
                  0.0
                                       1.0
                                                        1.0
18
                   0.0
                                       2.0
                                                        1.0
19
                   1.0
                                       0.0
                                                        0.0
```

20 0.0 3.0 1.0

1.4.4 Dealing with unknown categories

How does OneHotEncoder deal with unknown categories? Let's see an example:

```
[42]: X_toy = [['science', 10], ['arts', 30], ['arts', 20]]
      columns=['subject', 'group']
      pd.DataFrame(X_toy, columns=columns)
[42]:
         subject group
      0 science
      1
            arts
                     30
      2
            arts
                     20
[43]: ohe = OneHotEncoder(handle_unknown='error') # default value for handle_unknown
      ⇔is 'error'
      ohe.fit(X_toy);
[44]: columns_ohe = ohe.get_feature_names_out(['subject', 'group']).tolist()
      columns ohe
[44]: ['subject_arts', 'subject_science', 'group_10', 'group_20', 'group_30']
[45]: ohe.categories_
[45]: [array(['arts', 'science'], dtype=object), array([10, 20, 30], dtype=object)]
[46]: ex1 = ohe.transform([['arts', 10], ['science', 30]]).toarray()
      ex1
[46]: array([[1., 0., 1., 0., 0.],
             [0., 1., 0., 0., 1.]])
[47]: pd.DataFrame(ex1, columns=columns_ohe)
[47]:
         subject_arts subject_science group_10 group_20 group_30
      0
                  1.0
                                   0.0
                                             1.0
                                                       0.0
                                                                 0.0
      1
                  0.0
                                   1.0
                                             0.0
                                                       0.0
                                                                 1.0
[48]: # ex2 = ohe.transform([['arts', 10], ['science', 4]]).toarray()
      # This would give an error:
      # ValueError: Found unknown categories [4] in column 1 during transform
[49]: ohe = OneHotEncoder(handle_unknown='ignore') # now use 'ignore' instead of
       →'error'
      ohe.fit(X_toy);
```

```
[50]: ex2 = ohe.transform([['arts', 10], ['science', 4]]).toarray()
      ex2
[50]: array([[1., 0., 1., 0., 0.],
             [0., 1., 0., 0., 0.]
[51]: pd.DataFrame(ex2, columns=columns_ohe) # all "group" columns are 0 for value 4
[51]:
         subject_arts subject_science group_10 group_20
                                                              group_30
                  1.0
                                    0.0
                                               1.0
                                                         0.0
                                                                    0.0
                  0.0
                                    1.0
                                               0.0
                                                         0.0
                                                                   0.0
      1
[52]: ex3 = ohe.inverse_transform([[0, 1, 1, 0, 0], [1, 0, 0, 0, 0]])
      ex3
[52]: array([['science', 10],
             ['arts', None]], dtype=object)
[53]: pd.DataFrame(ex3, columns=columns)
[53]:
         subject group
      0
        science
                    10
      1
            arts
                  None
[54]: ex4 = ohe.inverse_transform([[0, 1, 0, 0, 1], [0, 0, 0, 1, 0]])
      ex4
[54]: array([['science', 30],
             [None, 20]], dtype=object)
[55]: pd.DataFrame(ex4, columns=columns)
[55]:
         subject group
         science
                    30
      0
      1
            None
                    20
     What if we know the possible categories beforehand? We can specify categories ahead of time.
[56]: ohe = OneHotEncoder(handle_unknown='error', categories=[['arts', 'science'],
      \rightarrow[10, 20, 30, 4]])
      ohe.fit(X_toy);
     Even though handle_unknown='error', ex2 does not give error anymore because categories are
     known.
[57]: ex2 = ohe.transform([['arts', 10], ['science', 4]]).toarray()
      ex2
```

```
[0., 1., 0., 0., 0., 1.]]
[58]: columns_ohe = ohe.get_feature_names_out(['subject', 'group']).tolist()
      pd.DataFrame(ex2, columns=columns_ohe)
[58]:
         subject_arts subject_science group_10 group_20 group_30
                                                                       group_4
      0
                  1.0
                                   0.0
                                              1.0
                                                        0.0
                                                                  0.0
                                                                            0.0
                  0.0
                                              0.0
                                                        0.0
                                                                  0.0
      1
                                   1.0
                                                                            1.0
     Dealing with unknown categories in cross_validate Let's create a pipeline with the column
     transformer and pass it to cross_validate.
[59]: ct
[59]: ColumnTransformer(transformers=[('pipeline',
                                       Pipeline(steps=[('simpleimputer',
                                                         SimpleImputer()),
                                                        ('standardscaler',
                                                         StandardScaler())]),
                                        ['university_years', 'lab1', 'lab2', 'lab3',
                                         'lab4', 'quiz1']),
                                       ('onehotencoder', OneHotEncoder(), ['major']),
                                       ('ordinalencoder',
                                       OrdinalEncoder(categories=[['Poor', 'Average',
                                                                     'Good',
                                                                     'Excellent']],
                                                       dtype=<class 'int'>),
                                        ['class_attendance']),
                                       ('passthrough', 'passthrough',
                                        ['ml_experience']),
                                       ('drop', 'drop', ['enjoy_course'])])
[60]: pipe = make_pipeline(ct, SVC())
[61]: | scores = cross_validate(pipe, X, y, return_train_score=True)
      pd.DataFrame(scores)
     /home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
     packages/sklearn/model_selection/_validation.py:770: UserWarning: Scoring
     failed. The score on this train-test partition for these parameters will be set
     to nan. Details:
     Traceback (most recent call last):
       File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
     packages/sklearn/model_selection/_validation.py", line 761, in _score
         scores = scorer(estimator, X_test, y_test)
       File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
     packages/sklearn/metrics/_scorer.py", line 418, in _passthrough_scorer
```

[57]: array([[1., 0., 1., 0., 0., 0.],

```
return estimator.score(*args, **kwargs)
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/sklearn/utils/metaestimators.py", line 113, in <lambda>
    out = lambda *args, **kwargs: self.fn(obj, *args, **kwargs) # noqa
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/sklearn/pipeline.py", line 707, in score
    Xt = transform.transform(Xt)
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/sklearn/compose/_column_transformer.py", line 748, in transform
    Xs = self._fit_transform(
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/sklearn/compose/_column_transformer.py", line 606, in _fit_transform
    return Parallel(n_jobs=self.n_jobs)(
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/joblib/parallel.py", line 1044, in __call__
    while self.dispatch_one_batch(iterator):
  File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/joblib/parallel.py", line 859, in dispatch_one_batch
    self._dispatch(tasks)
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/joblib/parallel.py", line 777, in _dispatch
    job = self. backend.apply async(batch, callback=cb)
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/joblib/_parallel_backends.py", line 208, in apply_async
    result = ImmediateResult(func)
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/joblib/_parallel_backends.py", line 572, in __init__
    self.results = batch()
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/joblib/parallel.py", line 262, in __call__
    return [func(*args, **kwargs)
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/joblib/parallel.py", line 262, in <listcomp>
    return [func(*args, **kwargs)
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/sklearn/utils/fixes.py", line 216, in __call__
   return self.function(*args, **kwargs)
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/sklearn/pipeline.py", line 876, in _transform_one
    res = transformer.transform(X)
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/sklearn/preprocessing/_encoders.py", line 509, in transform
    X_int, X_mask = self._transform(
 File "/home/moveisi/miniconda3/envs/cpsc330/lib/python3.10/site-
packages/sklearn/preprocessing/_encoders.py", line 142, in _transform
    raise ValueError(msg)
ValueError: Found unknown categories ['Biology'] in column 0 during transform
```

warnings.warn(

```
[61]:
         fit_time
                   score_time test_score train_score
      0 0.008816
                     0.005662
                                      1.00
                                               0.937500
      1 0.008196
                     0.004937
                                      1.00
                                               0.941176
      2 0.007402
                     0.003597
                                     0.50
                                               1.000000
      3 0.006360
                     0.004390
                                      0.75
                                               0.941176
      4 0.006452
                     0.006439
                                       NaN
                                               1.000000
```

- What's going on here??
- Let's look at the error message:
 - ValueError: Found unknown categories ['Biology'] in column 0 during transform
 - Same error that we got in the OneHotEncoder example above (ex2)

```
[62]: X["major"].value_counts()
```

```
[62]: Computer Science
                                 4
                                  4
      Mathematics
      Mechanical Engineering
                                  3
                                  3
      Psychology
      Economics
                                 2
                                  2
      Linguistics
                                  2
      Physics
      Biology
      Name: major, dtype: int64
```

- There is only one instance of Biology.
- During cross-validation, this is getting put into the validation split.
- By default, OneHotEncoder throws an error because you might want to know about this.

Simplest fix: - Pass handle_unknown="ignore" argument to OneHotEncoder - It creates a row with all zeros (as we saw in the example above)

```
("drop", drop_feats), # drop the drop features
      )
[64]: ct
[64]: ColumnTransformer(transformers=[('pipeline',
                                        Pipeline(steps=[('simpleimputer',
                                                         SimpleImputer()),
                                                         ('standardscaler',
                                                         StandardScaler())]),
                                        ['university_years', 'lab1', 'lab2', 'lab3',
                                         'lab4', 'quiz1']),
                                       ('onehotencoder',
                                        OneHotEncoder(handle_unknown='ignore'),
                                        ['major']),
                                       ('ordinalencoder',
                                        OrdinalEncoder(categories=[['Poor', 'Average',
                                                                     'Good',
                                                                     'Excellent']],
                                                       dtype=<class 'int'>),
                                        ['class_attendance']),
                                       ('passthrough', 'passthrough',
                                        ['ml_experience']),
                                       ('drop', 'drop', ['enjoy_course'])])
[65]: pipe = make_pipeline(ct, SVC())
[66]: | scores = cross_validate(pipe, X, y, cv=5, return_train_score=True)
      pd.DataFrame(scores)
[66]:
         fit_time
                   score_time
                               test_score
                                            train_score
      0 0.008285
                     0.006410
                                               0.937500
                                      1.00
      1 0.008545
                     0.004508
                                      1.00
                                               0.941176
      2 0.006607
                     0.003485
                                      0.50
                                               1.000000
      3 0.006461
                     0.005872
                                      0.75
                                               0.941176
      4 0.007575
                     0.004189
                                      0.75
                                               1.000000
```

• With this approach, all unknown categories will be represented with all zeros and cross-validation is running OK now.

Ask yourself the following questions when you work with categorical variables

- Do you want this behaviour? - Are you expecting to get many unknown categories? Do you want to be able to distinguish between them?

Cases where it's OK to break the golden rule We saw above that if we know categories beforehand we can specify them to OneHotEncoder to avoid errors during cross_over assessments. However, wouldn't that be *breaking the golden* rule?

When we know the categories in advance and this is one of the cases where it might be OK to violate the golden rule and get a list of all possible values for the categorical variable.

For example, if it's some fix number of categories. E.g., if it's something like: - provinces in Canada or - majors taught at UBC.

1.4.5 Categorical features with only two possible categories

- Sometimes you have features with only two possible categories.
- If we apply OheHotEncoder on such columns, it'll create two columns, which seems wasteful, as we could represent all information in the column in just one column with say 0's and 1's with presence of absence of one of one of the categories.
- You can pass drop="if_binary" argument to OneHotEncoder in order to create only one column in such scenario.

```
[67]: X["enjoy_course"].head()
[67]: 0
           ves
      2
           yes
      3
            no
      4
           yes
      Name: enjoy_course, dtype: object
[68]: ohe enc = OneHotEncoder(drop="if binary", dtype=int, sparse=False)
      ohe_enc.fit(X[["enjoy_course"]])
      transformed = ohe_enc.transform(X[["enjoy_course"]])
      df = pd.DataFrame(data=transformed, columns=["enjoy_course_enc"], index=X.index)
      X[["enjoy_course"]].join(df).head(10)
[68]:
        enjoy_course
                       enjoy_course_enc
      0
                                       1
                  yes
      1
                                       1
                  yes
      2
                                       1
                  yes
      3
                   no
                                       0
      4
                                       1
                  yes
      5
                                       0
                   no
      6
                                       1
                  yes
      7
                                       0
                   no
      8
                                       0
                   no
      9
                                       1
                  yes
[69]: numeric_feats = [
          "university_years",
          "lab1",
          "lab2",
           "lab3",
          "lab4",
           "quiz1",
```

```
] # apply scaling
      categorical_feats = ["major"] # apply one-hot encoding
      ordinal_feats = ["class_attendance"] # apply ordinal encoding
      binary_feats = ["enjoy_course"] # apply one-hot encoding with drop="if_binary"
      passthrough_feats = ["ml_experience"] # do not apply any transformation
      drop_feats = []
[70]: ct = make_column_transformer(
              make_pipeline(SimpleImputer(), StandardScaler()),
              numeric feats,
          ), # scaling on numeric features
              OneHotEncoder(handle_unknown="ignore"),
              categorical_feats,
          ), # OHE on categorical features
              OrdinalEncoder(categories=[class_attendance_levels], dtype=int),
              ordinal_feats,
          ), # Ordinal encoding on ordinal features
              OneHotEncoder(drop="if_binary", dtype=int),
              binary_feats,
          ), # OHE on categorical features
          ("passthrough", passthrough_feats), # no transformations on the binary_
       \hookrightarrow features
[71]: ct
[71]: ColumnTransformer(transformers=[('pipeline',
                                       Pipeline(steps=[('simpleimputer',
                                                         SimpleImputer()),
                                                        ('standardscaler',
                                                         StandardScaler())]),
                                        ['university_years', 'lab1', 'lab2', 'lab3',
                                         'lab4', 'quiz1']),
                                       ('onehotencoder-1',
                                       OneHotEncoder(handle_unknown='ignore'),
                                        ['major']),
                                      ('ordinalencoder',
                                       OrdinalEncoder(categories=[['Poor', 'Average',
                                                                    'Good',
                                                                    'Excellent']],
                                                       dtype=<class 'int'>),
                                        ['class attendance']),
                                      ('onehotencoder-2',
```

```
dtype=<class 'int'>),
                                        ['enjoy_course']),
                                       ('passthrough', 'passthrough',
                                        ['ml_experience'])])
[72]: pipe = make_pipeline(ct, SVC())
[73]: scores = cross_validate(pipe, X, y, cv=5, return_train_score=True)
      pd.DataFrame(scores)
[73]:
         fit_time
                   score_time
                               test_score
                                           train_score
      0 0.010822
                     0.005830
                                     1.00
                                               1.000000
      1 0.008248
                     0.004571
                                     1.00
                                               0.941176
      2 0.008762
                     0.005081
                                     0.50
                                               1.000000
      3 0.008046
                                     1.00
                     0.005495
                                               0.941176
      4 0.007485
                                     0.75
                                               1.000000
                     0.004781
```

OneHotEncoder(drop='if_binary',

Note Do not read too much into the scores, as we are running cross-validation on a very small dataset with 21 examples. The main point here is to show you how can we use ColumnTransformer to apply different transformations on different columns.

1.5 Break (5 min)



1.6 ColumnTransformer on the California housing dataset

[74]:		longitude	latitude	housing_median_age	total_rooms	total_bedrooms	\
	6051	-117.75	34.04	22.0	2948.0	636.0	
	20113	-119.57	37.94	17.0	346.0	130.0	
	14289	-117.13	32.74	46.0	3355.0	768.0	
	13665	-117.31	34.02	18.0	1634.0	274.0	
	14471	-117.23	32.88	18.0	5566.0	1465.0	

```
population households median_income median_house_value \
6051
           2600.0
                         602.0
                                       3.1250
                                                          113600.0
                          20.0
20113
             51.0
                                       3.4861
                                                          137500.0
14289
           1457.0
                         708.0
                                       2.6604
                                                          170100.0
13665
            899.0
                         285.0
                                       5.2139
                                                          129300.0
14471
           6303.0
                        1458.0
                                       1.8580
                                                          205000.0
      ocean_proximity
6051
               INLAND
20113
               INLAND
14289
           NEAR OCEAN
13665
               INLAND
14471
           NEAR OCEAN
```

Some column values are mean/median but some are not.

Let's add some new features to the dataset which could help predicting the target: median_house_value.

[76]: train_df.head()

```
[76]:
             longitude
                        latitude
                                  housing median age total rooms
                                                                    total bedrooms
      6051
               -117.75
                           34.04
                                                 22.0
                                                            2948.0
                                                                             636.0
      20113
               -119.57
                           37.94
                                                 17.0
                                                             346.0
                                                                             130.0
      14289
              -117.13
                           32.74
                                                 46.0
                                                            3355.0
                                                                             768.0
                           34.02
                                                 18.0
      13665
              -117.31
                                                            1634.0
                                                                             274.0
      14471
              -117.23
                           32.88
                                                 18.0
                                                            5566.0
                                                                            1465.0
```

```
6051
                 2600.0
                               602.0
                                              3.1250
                                                                 113600.0
                                20.0
      20113
                    51.0
                                              3.4861
                                                                 137500.0
      14289
                 1457.0
                               708.0
                                              2.6604
                                                                 170100.0
      13665
                  899.0
                               285.0
                                              5.2139
                                                                 129300.0
      14471
                 6303.0
                              1458.0
                                              1.8580
                                                                 205000.0
            ocean_proximity
                              rooms per household
                                                    bedrooms per household \
      6051
                      INLAND
                                          4.897010
                                                                   1.056478
      20113
                      INLAND
                                         17.300000
                                                                   6.500000
      14289
                 NEAR OCEAN
                                          4.738701
                                                                   1.084746
      13665
                      INLAND
                                          5.733333
                                                                   0.961404
      14471
                 NEAR OCEAN
                                          3.817558
                                                                   1.004801
             population_per_household
      6051
                              4.318937
      20113
                              2.550000
      14289
                              2.057910
      13665
                              3.154386
      14471
                              4.323045
[77]: # Let's keep both numeric and categorical columns in the data.
      X_train = train_df.drop(columns=["median_house_value"])
      y_train = train_df["median_house_value"]
      X_test = test_df.drop(columns=["median_house_value"])
      y_test = test_df["median_house_value"]
[78]: from sklearn.compose import ColumnTransformer, make_column_transformer
[79]: X_train.head(10)
[79]:
             longitude
                         latitude
                                   housing_median_age
                                                        total_rooms
                                                                      total_bedrooms
      6051
               -117.75
                            34.04
                                                  22.0
                                                              2948.0
                                                                                636.0
      20113
               -119.57
                            37.94
                                                  17.0
                                                               346.0
                                                                                130.0
                            32.74
      14289
               -117.13
                                                  46.0
                                                              3355.0
                                                                                768.0
      13665
               -117.31
                            34.02
                                                  18.0
                                                              1634.0
                                                                                274.0
      14471
               -117.23
                            32.88
                                                  18.0
                                                              5566.0
                                                                               1465.0
      9730
               -121.74
                            36.79
                                                  16.0
                                                              3841.0
                                                                                620.0
      14690
               -117.09
                            32.80
                                                  36.0
                                                              2163.0
                                                                                367.0
                                                                                410.0
      7938
               -118.11
                            33.86
                                                  33.0
                                                              2389.0
      18365
               -122.12
                            37.28
                                                  21.0
                                                               349.0
                                                                                 64.0
               -117.91
                            33.74
                                                  25.0
                                                                                965.0
      10931
                                                              4273.0
             population households
                                      median_income ocean_proximity \
      6051
                 2600.0
                               602.0
                                              3.1250
                                                               INLAND
```

median_income median_house_value

population households

```
20113
                   51.0
                                20.0
                                             3.4861
                                                              INLAND
      14289
                 1457.0
                               708.0
                                             2.6604
                                                          NEAR OCEAN
      13665
                  899.0
                               285.0
                                             5.2139
                                                              INLAND
      14471
                 6303.0
                              1458.0
                                             1.8580
                                                          NEAR OCEAN
      9730
                 1799.0
                               611.0
                                             4.3814
                                                           <1H OCEAN
      14690
                  915.0
                               360.0
                                             4.7188
                                                          NEAR OCEAN
      7938
                 1229.0
                               393.0
                                             5.3889
                                                           <1H OCEAN
      18365
                  149.0
                                56.0
                                             5.8691
                                                           <1H OCEAN
      10931
                               922.0
                                             2.9926
                                                           <1H OCEAN
                 2946.0
             rooms_per_household bedrooms_per_household population_per_household
      6051
                        4.897010
                                                  1.056478
                                                                             4.318937
      20113
                        17.300000
                                                  6.500000
                                                                             2.550000
      14289
                        4.738701
                                                  1.084746
                                                                             2.057910
      13665
                         5.733333
                                                  0.961404
                                                                             3.154386
      14471
                        3.817558
                                                  1.004801
                                                                             4.323045
      9730
                         6.286416
                                                                             2.944354
                                                  1.014730
                         6.008333
      14690
                                                  1.019444
                                                                             2.541667
      7938
                         6.078880
                                                  1.043257
                                                                             3.127226
      18365
                         6.232143
                                                  1.142857
                                                                             2.660714
      10931
                         4.634490
                                                                             3.195228
                                                  1.046638
[80]: X train.columns
[80]: Index(['longitude', 'latitude', 'housing median age', 'total rooms',
             'total_bedrooms', 'population', 'households', 'median_income',
             'ocean_proximity', 'rooms_per_household', 'bedrooms_per_household',
             'population_per_household'],
            dtype='object')
[81]: # Identify the categorical and numeric columns
      numeric_features = [
          "longitude",
          "latitude",
          "housing_median_age",
          "total rooms",
          "total bedrooms",
          "population",
          "households",
          "median_income",
          "rooms_per_household",
          "bedrooms_per_household",
          "population_per_household",
      ]
      categorical_features = ["ocean_proximity"]
      target = "median_house_value"
```

• Let's create a ColumnTransformer for our dataset.

```
[82]: X_train.info()
     <class 'pandas.core.frame.DataFrame'>
     Int64Index: 18576 entries, 6051 to 19966
     Data columns (total 12 columns):
          Column
                                    Non-Null Count Dtype
                                    18576 non-null float64
      0
          longitude
      1
          latitude
                                    18576 non-null float64
         housing_median_age
                                    18576 non-null float64
      3
         total_rooms
                                    18576 non-null float64
      4
          total_bedrooms
                                    18391 non-null float64
      5
          population
                                    18576 non-null float64
      6
          households
                                    18576 non-null float64
      7
          median_income
                                    18576 non-null float64
      8
          ocean_proximity
                                    18576 non-null object
          rooms per household
                                    18576 non-null float64
      10 bedrooms_per_household
                                    18391 non-null float64
      11 population_per_household 18576 non-null float64
     dtypes: float64(11), object(1)
     memory usage: 1.8+ MB
[83]: X_train["ocean_proximity"].value_counts()
[83]: <1H OCEAN
                    8221
      INLAND
                    5915
      NEAR OCEAN
                    2389
      NEAR BAY
                    2046
      ISLAND
     Name: ocean_proximity, dtype: int64
[84]: numeric_transformer = make_pipeline(SimpleImputer(strategy="median"),__

→StandardScaler())
      categorical_transformer = OneHotEncoder(handle_unknown="ignore")
      preprocessor = make_column_transformer(
          (numeric_transformer, numeric_features),
          (categorical_transformer, categorical_features),
[85]: preprocessor
[85]: ColumnTransformer(transformers=[('pipeline',
                                       Pipeline(steps=[('simpleimputer',
      SimpleImputer(strategy='median')),
                                                       ('standardscaler',
```

```
StandardScaler())]),
['longitude', 'latitude', 'housing_median_age',
   'total_rooms', 'total_bedrooms', 'population',
   'households', 'median_income',
   'rooms_per_household',
   'bedrooms_per_household',
   'population_per_household']),
('onehotencoder',
OneHotEncoder(handle_unknown='ignore'),
['ocean_proximity'])])
```

```
[86]: X_train_pp = preprocessor.fit_transform(X_train)
```

- When we fit the preprocessor, it calls fit on *all* the transformers
- When we transform the preprocessor, it calls transform on all the transformers.

We can get the new names of the columns that were generated by the one-hot encoding:

```
[87]: preprocessor
[87]: ColumnTransformer(transformers=[('pipeline',
                                        Pipeline(steps=[('simpleimputer',
      SimpleImputer(strategy='median')),
                                                        ('standardscaler',
                                                         StandardScaler())]),
                                        ['longitude', 'latitude', 'housing median age',
                                         'total_rooms', 'total_bedrooms', 'population',
                                         'households', 'median income',
                                         'rooms_per_household',
                                         'bedrooms per household',
                                         'population_per_household']),
                                       ('onehotencoder',
                                        OneHotEncoder(handle_unknown='ignore'),
                                        ['ocean_proximity'])])
[88]: preprocessor.named_transformers_["onehotencoder"].get_feature_names_out(
          categorical features
      )
[88]: array(['ocean_proximity_<1H_OCEAN', 'ocean_proximity_INLAND',
             'ocean_proximity_ISLAND', 'ocean_proximity_NEAR BAY',
             'ocean_proximity_NEAR OCEAN'], dtype=object)
     Combining this with the numeric feature names gives us all the column names:
[89]: column_names = numeric_features + list(
          preprocessor.named_transformers_["onehotencoder"].get_feature_names_out(
              categorical_features
          )
```

```
column_names
[89]: ['longitude',
       'latitude',
       'housing_median_age',
       'total rooms',
       'total_bedrooms',
       'population',
       'households',
       'median_income',
       'rooms_per_household',
       'bedrooms_per_household',
       'population_per_household',
       'ocean_proximity_<1H OCEAN',
       'ocean_proximity_INLAND',
       'ocean_proximity_ISLAND',
       'ocean_proximity_NEAR BAY',
       'ocean_proximity_NEAR OCEAN']
     Let's visualize the preprocessed training data as a dataframe.
[90]: pd.DataFrame(X_train_pp, columns=column_names)
[90]:
             longitude latitude
                                   housing_median_age
                                                        total_rooms
                                                                      total_bedrooms
      0
              0.908140 -0.743917
                                             -0.526078
                                                            0.143120
                                                                             0.235339
      1
             -0.002057 1.083123
                                             -0.923283
                                                           -1.049510
                                                                            -0.969959
      2
              1.218207 -1.352930
                                              1.380504
                                                            0.329670
                                                                             0.549764
      3
              1.128188 -0.753286
                                             -0.843842
                                                           -0.459154
                                                                            -0.626949
      4
              1.168196 -1.287344
                                             -0.843842
                                                                             2.210026
                                                            1.343085
      18571
              0.733102 -0.804818
                                              0.586095
                                                           -0.875337
                                                                            -0.243446
      18572
              1.163195 -1.057793
                                             -1.161606
                                                            0.940194
                                                                             0.609314
      18573
             -1.097293 0.797355
                                             -1.876574
                                                            0.695434
                                                                             0.433046
      18574
             -1.437367
                         1.008167
                                              1.221622
                                                           -0.499947
                                                                            -0.484029
      18575
              0.242996 0.272667
                                             -0.684960
                                                           -0.332190
                                                                            -0.353018
             population
                         households
                                      median_income
                                                      rooms_per_household
      0
                            0.266135
               1.026092
                                           -0.389736
                                                                 -0.210591
      1
              -1.206672
                           -1.253312
                                           -0.198924
                                                                  4.726412
      2
               0.024896
                            0.542873
                                           -0.635239
                                                                 -0.273606
      3
              -0.463877
                           -0.561467
                                            0.714077
                                                                  0.122307
      4
               4.269688
                            2.500924
                                           -1.059242
                                                                 -0.640266
      18571
              -0.822136
                           -0.966131
                                           -0.118182
                                                                  0.063110
               0.882438
                                            0.357500
      18572
                            0.728235
                                                                  0.235096
      18573
               0.881563
                            0.514155
                                            0.934269
                                                                  0.211892
```

0.006578

-0.273382

18574

-0.759944

-0.454427

```
18575
                     -0.396991
                                                             0.025998
        -0.164307
                                     -0.711754
                                 population_per_household
       bedrooms_per_household
0
                     -0.083813
                                                   0.126398
1
                     11.166631
                                                  -0.050132
2
                     -0.025391
                                                  -0.099240
3
                     -0.280310
                                                  0.010183
4
                     -0.190617
                                                  0.126808
18571
                     -0.099558
                                                  0.071541
18572
                     -0.163397
                                                  0.007458
18573
                     -0.135305
                                                  0.044029
18574
                     -0.149822
                                                  -0.132875
18575
                      0.042957
                                                  0.051269
       ocean_proximity_<1H OCEAN ocean_proximity_INLAND
0
                               0.0
                                                         1.0
                               0.0
1
                                                         1.0
2
                               0.0
                                                         0.0
3
                               0.0
                                                         1.0
4
                               0.0
                                                         0.0
18571
                               1.0
                                                         0.0
                               1.0
                                                         0.0
18572
                               1.0
18573
                                                         0.0
                               0.0
18574
                                                         0.0
18575
                               0.0
                                                         1.0
       ocean_proximity_ISLAND
                                ocean_proximity_NEAR BAY \
0
                            0.0
                                                        0.0
1
                            0.0
                                                        0.0
2
                            0.0
                                                        0.0
3
                            0.0
                                                        0.0
                                                        0.0
4
                            0.0
18571
                            0.0
                                                        0.0
                            0.0
                                                        0.0
18572
18573
                            0.0
                                                        0.0
                            0.0
                                                        1.0
18574
18575
                            0.0
                                                        0.0
       {\tt ocean\_proximity\_NEAR~OCEAN}
0
                                0.0
                                0.0
1
2
                                1.0
3
                                0.0
4
```

1.0

```
18571
                                    0.0
      18572
                                    0.0
      18573
                                    0.0
      18574
                                    0.0
      18575
                                    0.0
      [18576 rows x 16 columns]
[91]: y_train.to_frame().head()
[91]:
             median_house_value
      6051
                       113600.0
      20113
                       137500.0
      14289
                       170100.0
      13665
                       129300.0
      14471
                       205000.0
[92]: results_dict = {}
      dummy = DummyRegressor()
      results_dict["dummy"] = mean_std_cross_val_scores(
          dummy, X_train, y_train, return_train_score=True
      pd.DataFrame(results_dict).T
[92]:
                      fit_time
                                       score_time
                                                            test_score \
      dummy 0.001 (+/-0.000) 0.000 (+/-0.000) -0.001 (+/-0.001)
                   train_score
      dummy 0.000 (+/-0.000)
[93]: from sklearn.svm import SVR
      knn_pipe = make_pipeline(preprocessor, KNeighborsRegressor())
[94]: knn_pipe
[94]: Pipeline(steps=[('columntransformer',
                       ColumnTransformer(transformers=[('pipeline',
      Pipeline(steps=[('simpleimputer',
      SimpleImputer(strategy='median')),
      ('standardscaler',
      StandardScaler())]),
                                                         ['longitude', 'latitude',
                                                          'housing_median_age',
                                                          'total_rooms',
                                                          'total_bedrooms',
```

```
'population', 'households',
                                                          'median_income',
                                                          'rooms_per_household',
                                                         'bedrooms_per_household',
                                                          'population_per_household']),
                                                       ('onehotencoder',
      OneHotEncoder(handle_unknown='ignore'),
                                                        ['ocean_proximity'])])),
                      ('kneighborsregressor', KNeighborsRegressor())])
[95]: results_dict["imp + scaling + ohe + KNN"] = mean_std_cross_val_scores(
          knn_pipe, X_train, y_train, return_train_score=True
[96]: pd.DataFrame(results_dict).T
[96]:
                                          fit_time
                                                           score_time \
                                 0.001 (+/- 0.000) 0.000 (+/- 0.000)
      dummy
      imp + scaling + ohe + KNN 0.024 (+/- 0.002) 0.807 (+/- 0.136)
                                         test_score
                                                           train_score
                                 -0.001 (+/- 0.001) 0.000 (+/- 0.000)
      dummy
                                  0.721 (+/- 0.012) 0.816 (+/- 0.006)
      imp + scaling + ohe + KNN
[97]: svr_pipe = make_pipeline(preprocessor, SVR())
      results_dict["imp + scaling + ohe + SVR (default)"] = mean_std_cross_val_scores(
          svr_pipe, X_train, y_train, return_train_score=True
[98]: pd.DataFrame(results_dict).T
[98]:
                                                     fit_time
                                                                      score_time \
                                            0.001 (+/- 0.000) 0.000 (+/- 0.000)
      dummy
                                            0.024 (+/- 0.002) 0.807 (+/- 0.136)
      imp + scaling + ohe + KNN
      imp + scaling + ohe + SVR (default) 12.143 (+/- 1.702) 3.046 (+/- 0.437)
                                                   test_score
                                                                      train_score
      dummy
                                           -0.001 (+/- 0.001)
                                                                0.000 (+/- 0.000)
                                            0.721 (+/- 0.012)
                                                                0.816 (+/- 0.006)
      imp + scaling + ohe + KNN
      imp + scaling + ohe + SVR (default) -0.049 (+/- 0.012) -0.049 (+/- 0.001)
     The results with scikit-learn's default SVR hyperparameters are pretty bad.
[99]: svr_C_pipe = make_pipeline(preprocessor, SVR(C=10000))
      results_dict["imp + scaling + ohe + SVR (C=10000)"] = mean_std_cross_val_scores(
          svr_C_pipe, X_train, y_train, return_train_score=True
      )
```

[100]: pd.DataFrame(results_dict).T 「100]: fit time score_time \ 0.000 (+/- 0.000)dummy 0.001 (+/- 0.000) 0.024 (+/- 0.002) imp + scaling + ohe + KNN 0.807 (+/- 0.136)imp + scaling + ohe + SVR (default) 12.143 (+/-1.702)3.046 (+/- 0.437) imp + scaling + ohe + SVR (C=10000) 11.154 (+/- 1.224) 3.165 (+/- 0.114)test_score train_score dummy -0.001 (+/- 0.001) 0.000 (+/- 0.000)0.721 (+/- 0.012) 0.816 (+/- 0.006) imp + scaling + ohe + KNN -0.049 (+/- 0.012)-0.049 (+/- 0.001) imp + scaling + ohe + SVR (default) imp + scaling + ohe + SVR (C=10000) 0.721 (+/- 0.007) 0.726 (+/-0.007)

With a bigger value for C the results are much better. We need to carry out systematic hyper-parameter optimization to get better results. (Coming up next week.)

• Note that categorical features are different than free text features. Sometimes there are columns containing free text information and we we'll look at ways to deal with them in the later part of this lecture.

1.6.1 OHE with many categories

- Do we have enough data for rare categories to learn anything meaningful?
- How about grouping them into bigger categories?
 - Example: country names into continents such as "South America" or "Asia"
- Or having "other" category for rare cases?

1.6.2 Do we actually want to use certain features for prediction?

- Do you *want* to use certain features such as **gender** or **race** in prediction?
- Remember that the systems you build are going to be used in some applications.
- It's extremely important to be mindful of the consequences of including certain features in your predictive model.

1.6.3 Preprocessing the targets?

- Generally no need for this when doing classification.
- In regression it makes sense in some cases. More on this later.
- sklearn is fine with categorical labels (y-values) for classification problems.

1.6.4 Questions for you

True/False: Categorical features

- 1. handle_unknown="ignore" would treat all unknown categories equally.
- 2. Creating groups of rarely occurring categories might overfit the model.

1.7 Encoding text data

```
[101]: toy_spam = [
              "URGENT!! As a valued network customer you have been selected to \sqcup
       ⇒receive a £900 prize reward!",
             "spam",
          ],
          ["Lol you are always so convincing.", "non spam"],
          ["Nah I don't think he goes to usf, he lives around here though", "non_
       ⇔spam"],
          "URGENT! You have won a 1 week FREE membership in our £100000 prize_

    Jackpot!",

              "spam",
          ],
              ⇔latest colour mobiles with camera for Free! Call The Mobile Update Co FREE⊔
       on 08002986030",
             "spam",
          ],
          ["Congrats! I can't wait to see you!!", "non spam"],
      toy_df = pd.DataFrame(toy_spam, columns=["sms", "target"])
```

1.7.1 Spam/non spam toy example

- What if the feature is in the form of raw text?
- The feature sms below is neither categorical nor ordinal.
- How can we encode it so that we can pass it to the machine learning algorithms we have seen so far?

```
[102]: toy_df.style.set_properties(**{"text-align": "left"})
```

[102]: <pandas.io.formats.style.Styler at 0x7f7caae7fcd0>

1.7.2 What if we apply OHE?

```
[103]: ### DO NOT DO THIS.
enc = OneHotEncoder(sparse=False)
    transformed = enc.fit_transform(toy_df[["sms"]])
    pd.DataFrame(transformed, columns=enc.categories_)
```

```
0.0
3
4
                                   0.0
5
                                   1.0
  Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030 \
0.0
0.0
0.0
0.0
1.0
5
0.0
  Lol you are always so convincing.
                                 0.0
                                 1.0
1
2
                                 0.0
                                 0.0
3
4
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  Nah I don't think he goes to usf, he lives around here though \
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  URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot! \
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  URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
1.0
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1 0.0 2 0.0 3 0.0 4 0.0 5

- We do **not have a fixed number** of categories here.
- Each "category" (feature value) is likely to **occur only once** in the training data and we won't learn anything meaningful if we apply one-hot encoding or ordinal encoding on this feature.
- How can we encode or represent **raw text data into fixed number of features** so that we can learn some useful patterns from it?
- This is a well studied problem in the field of *Natural Language Processing (NLP)*, which is concerned with giving computers the ability to understand written and spoken language.
- Some popular representations of raw text include:
 - Bag of words
 - TF-IDF
 - Embedding representations

1.7.3 Bag of words (BOW) representation

- One of the most popular representation of raw text
- Ignores the syntax and word order
- It has two components:
 - The vocabulary (all unique words in all documents)
 - A value indicating either the presence or absence or the count of each word in the document.

Source

1.7.4 Extracting BOW features using scikit-learn

- CountVectorizer
 - Converts a collection of text documents to a matrix of word counts.
 - Each row represents a "document" (e.g., a text message in our example).
 - Each column represents a word in the vocabulary (the set of unique words) in the training data.
 - Each cell represents how often the word occurs in the document.

Note In the Natural Language Processing (NLP) community text data is referred to as a **corpus** (plural: corpora).

```
[104]: from sklearn.feature_extraction.text import CountVectorizer
       vec = CountVectorizer()
       X_counts = vec.fit_transform(toy_df["sms"])
       bow_df = pd.DataFrame(
          X_counts.toarray(), columns=vec.get_feature_names_out(), index=toy_df["sms"]
       bow_df
[104]: 08002986030 \
      URGENT!! As a valued network customer you have been selected to receive a £900
      prize reward!
      Lol you are always so convincing.
      Nah I don't think he goes to usf, he lives around here though
      URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
      Had your mobile 11 months or more? U R entitled to Update to the latest colour
      mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
       Congrats! I can't wait to see you!!
       100000 \
      URGENT!! As a valued network customer you have been selected to receive a £900
      prize reward!
      Lol you are always so convincing.
      Nah I don't think he goes to usf, he lives around here though
      URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
      Had your mobile 11 months or more? U R entitled to Update to the latest colour
      mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
       Congrats! I can't wait to see you!!
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      URGENT!! As a valued network customer you have been selected to receive a £900
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Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
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Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
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Congrats! I can't wait to see you!!
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Nah I don't think he goes to usf, he lives around here though
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Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
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Congrats! I can't wait to see you!!
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mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
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Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
[6 rows x 61 columns]
1.7.5 Input to CountVectorizer.fit_transform
```

```
[105]: type(toy_df["sms"])
```

[105]: pandas.core.series.Series

Important Note that unlike other transformers we are **passing a Series** object to fit_transform. For other transformers, you can define one transformer for more than one columns. But with CountVectorizer you need to define **separate CountVectorizer transformers for each text column**, if you have more than one text columns.

1.7.6 Output of CountVectorizer.fit_transform

fit_transform has returned a sparse matrix:

```
[106]: X_counts
```

```
[106]: <6x61 sparse matrix of type '<class 'numpy.int64'>'
with 71 stored elements in Compressed Sparse Row format>
```

Why sparse matrices?

- Most words do not appear in a given document.
- We get massive computational savings if we only store the nonzero elements.
- $\bullet\,$ There is a bit of overhead, because we also need to store the locations:
 - e.g. "location (3,27): 1".
- However, if the fraction of nonzero is small, this is a huge win.

```
The number of rows and columns: 6 61
The total number of elements: 366
The number of non-zero elements: 71
Proportion of non-zero elements: 0.1940
The value at cell (4,51) is: 2
```

Question for you - What would happen if you apply StandardScaler on sparse data?

1.7.7 OneHotEncoder and sparse features

- By default, OneHotEncoder also creates sparse features.
- You could set sparse=False to get a regular numpy array.
- If there are a huge number of categories, it may be beneficial to keep them sparse.
- For smaller number of categories, it doesn't matter much.

1.7.8 Important hyperparameters of CountVectorizer

- binary
 - whether to use absence/presence feature values or counts
- max_features
 - only consider top max_features ordered by frequency in the corpus
- max_df

- max document frequency, ignore features which occur in more than max_df documents
- min_df
 - min document frequency, ignore features which occur in less than min_df documents
- ngram_range
 - consider word sequences in the given range

Let's look at all features, i.e., words (along with their frequencies).

```
[108]: vec = CountVectorizer()
       X_counts = vec.fit_transform(toy_df["sms"])
       bow_df = pd.DataFrame(
          X_counts.toarray(), columns=vec.get_feature_names_out(), index=toy_df["sms"]
       print("Max value: ", bow_df.max().max())
       bow_df
      Max value: 2
[108]: 08002986030 \
      URGENT!! As a valued network customer you have been selected to receive a £900
      prize reward!
      Lol you are always so convincing.
      Nah I don't think he goes to usf, he lives around here though
      URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
      Had your mobile 11 months or more? U R entitled to Update to the latest colour
      mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
       Congrats! I can't wait to see you!!
       100000 \
      URGENT!! As a valued network customer you have been selected to receive a £900
      prize reward!
      Lol you are always so convincing.
      Nah I don't think he goes to usf, he lives around here though
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Congrats! I can't wait to see you!!
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URGENT!! As a valued network customer you have been selected to receive a £900
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Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
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Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
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Congrats! I can't wait to see you!!
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Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
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Had your mobile 11 months or more? U R entitled to Update to the latest colour
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Congrats! I can't wait to see you!!
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mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
[6 rows x 61 columns]
```

When we use binary=True, the representation uses presence/absence of words instead of word counts.

```
[109]: vec_binary = CountVectorizer(binary=True)
       X_counts = vec_binary.fit_transform(toy_df["sms"])
       bow_df = pd.DataFrame(
          X_counts.toarray(), columns=vec_binary.get_feature_names_out(),__

index=toy_df["sms"]
       print("Max value: ", bow_df.max().max())
       bow df
      Max value: 1
[109]: 08002986030 \
      URGENT!! As a valued network customer you have been selected to receive a £900
      prize reward!
      Lol you are always so convincing.
      Nah I don't think he goes to usf, he lives around here though
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       Congrats! I can't wait to see you!!
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Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
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Congrats! I can't wait to see you!!
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Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
always \
URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
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Congrats! I can't wait to see you!!
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Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
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URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
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Had your mobile 11 months or more? U R entitled to Update to the latest colour mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030

Congrats! I can't wait to see you!!

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URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
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Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
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prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
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URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030 ...
Congrats! I can't wait to see you!!
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URGENT!! As a valued network customer you have been selected to receive a £900
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Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
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Congrats! I can't wait to see you!!
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1
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
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URGENT!! As a valued network customer you have been selected to receive a £900
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Lol you are always so convincing.
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Congrats! I can't wait to see you!!
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Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
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URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
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URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
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URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
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URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
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mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
       Congrats! I can't wait to see you!!
                                                                                    you
       \
       sms
      URGENT!! As a valued network customer you have been selected to receive a £900
      prize reward!
      Lol you are always so convincing.
      Nah I don't think he goes to usf, he lives around here though
      URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
      Had your mobile 11 months or more? U R entitled to Update to the latest colour
      mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
       Congrats! I can't wait to see you!!
                                                                                    your
       SMS
      URGENT!! As a valued network customer you have been selected to receive a £900
      prize reward!
      0
      Lol you are always so convincing.
      Nah I don't think he goes to usf, he lives around here though
      URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
      Had your mobile 11 months or more? U R entitled to Update to the latest colour
      mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
       Congrats! I can't wait to see you!!
       [6 rows x 61 columns]
      We can control the size of X (the number of features) using max_features.
[110]: vec8 = CountVectorizer(max_features=8)
       X_counts = vec8.fit_transform(toy_df["sms"])
       bow_df = pd.DataFrame(
           X_counts.toarray(), columns=vec8.get_feature_names_out(),_

index=toy_df["sms"]

       )
```

```
print("Max value: ", bow_df.max().max())
       print(bow_df.max())
       bow_df
      Max value: 2
      free
      have
                1
      mobile
                2
                2
      the
      update
                2
      urgent
                1
                1
      you
      dtype: int64
[110]:
                                                                                    free
       sms
      URGENT!! As a valued network customer you have been selected to receive a £900
      prize reward!
      0
      Lol you are always so convincing.
      Nah I don't think he goes to usf, he lives around here though
      URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
      Had your mobile 11 months or more? U R entitled to Update to the latest colour
      mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
       Congrats! I can't wait to see you!!
                                                                                    have
       \
      URGENT!! As a valued network customer you have been selected to receive a £900
      prize reward!
      Lol you are always so convincing.
      Nah I don't think he goes to usf, he lives around here though
      URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
      Had your mobile 11 months or more? U R entitled to Update to the latest colour
      mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
```

```
Congrats! I can't wait to see you!!
mobile \
URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
                                                                             the
\
sms
URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
                                                                             to
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URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
```

```
0
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
update \
sms
URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
urgent \
URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!
Had your mobile 11 months or more? U R entitled to Update to the latest colour
mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030
Congrats! I can't wait to see you!!
                                                                            you
URGENT!! As a valued network customer you have been selected to receive a £900
prize reward!
Lol you are always so convincing.
Nah I don't think he goes to usf, he lives around here though
```

```
URGENT! You have won a 1 week FREE membership in our £100000 prize Jackpot!

Had your mobile 11 months or more? U R entitled to Update to the latest colour mobiles with camera for Free! Call The Mobile Update Co FREE on 08002986030 (Congrats! I can't wait to see you!!
```

Note (Optional)

Notice that vec8 and vec8_binary have different vocabularies, which is kind of unexpected behaviour and doesn't match the documentation of scikit-learn.

Here is the code for binary=True condition in scikit-learn. As we can see, the binarization is done before limiting the features to max_features, and so now we are actually looking at the document counts (in how many documents it occurs) rather than term count. This is not explained anywhere in the documentation.

The ties in counts between different words makes it even more confusing. I don't think it'll have a big impact on the results but this is good to know! Remember that scikit-learn developers are also humans who are prone to make mistakes. So it's always a good habit to question whatever tools we use every now and then.

```
[111]:
                counts
       to
                      5
                      4
       you
       free
                      3
       have
                      2
       mobile
                      2
       the
                      2
       update
       urgent
```

```
[112]:
                  counts
        to
                        4
                        4
        you
                        2
       free
                        2
       have
                        2
       prize
       urgent
                        2
       mobiles
                        1
       months
                        1
```

1.7.9 Preprocessing

- Note that CountVectorizer is carrying out some preprocessing such as the following because of the default argument values:
 - Converting words to lowercase (lowercase=True)
 - getting rid of punctuation and special characters (token_pattern = '(?u)\\b\\\w\\\\b')

```
[113]: pipe = make_pipeline(CountVectorizer(), SVC())
[114]: pipe.fit(toy_df["sms"], toy_df["target"])
[114]: Pipeline(steps=[('countvectorizer', CountVectorizer()), ('svc', SVC())])
[115]: pipe.predict(toy_df["sms"]).tolist()
[115]: ['spam', 'non spam', 'non spam', 'spam', 'spam', 'non spam']
[116]: toy_df["target"].tolist()
[116]: ['spam', 'non spam', 'non spam', 'spam', 'spam', 'non spam']
```

1.7.10 Is this a realistic representation of text data?

- Of course this is not a great representation of language
 - We are throwing out everything we know about language and losing a lot of information.
 - It assumes that there is **no syntax and compositional meaning** in language.
- But it works surprisingly well for many tasks.
- $\bullet~$ We will learn more expressive representations in the coming weeks.

1.7.11 Questions for you

CountVectorizer: True or False

- 1. As you increase the value for max_features hyperparameter of CountVectorizer the training score is likely to go up.
- 2. Suppose you are encoding text data using CountVectorizer. If you encounter a word in the validation or the test split that's not available in the training data, we'll get an error.

- 3. max_df hyperparameter of CountVectorizer can be used to get rid of most frequently occurring words from the dictionary.
- 4. In the code below, inside cross_validate, each fold might have slightly different number of features (columns) in the fold.

```
pipe = (CountVectorizer(), SVC())
cross_validate(pipe, X_train, y_train)
```

4

5

6

7

8

Identify column transformations Consider the restaurant data from the survey you did a few weeks ago.

```
[117]: restaurant_data = pd.read_csv("data/cleaned_restaurant_data.csv")
       restaurant_data.head(10)
[117]:
         north_america
                          eat_out_freq
                                          age
                                               n_people
                                                            price
                                                                             food_type
                                                                                  Other
       0
                     Yes
                                    3.0
                                           29
                                                      20
                                                                10
       1
                     Yes
                                    2.0
                                           23
                                                      10
                                                                20
                                                                               Chinese
       2
                     Yes
                                    2.0
                                                      10
                                                                40
                                                                                  Other
                                           21
       3
                      No
                                    2.0
                                           24
                                                      15
                                                                40
                                                                                  Other
       4
                     Yes
                                    5.0
                                           23
                                                      20
                                                                10
                                                                    Canadian/American
       5
                     Yes
                                    2.0
                                           22
                                                      60
                                                                20
                                                                               Chinese
       6
                                    2.0
                                                          1000000
                                                                    Canadian/American
                     Yes
                                           23
                                               10000000
       7
                     Yes
                                    4.0
                                           20
                                                      20
                                                                40
                                                                               Chinese
                                                                60
                                                                               Italian
       8
                     Yes
                                    1.0
                                           21
                                                      40
                      No
                                    2.0
                                           20
                                                      50
                                                                45
                                                                    Canadian/American
         noise_level good_server
             no music
       0
                               Yes
       1
                   low
                                Yes
       2
                   low
                                Yes
       3
               medium
                                Yes
       4
                  low
                                Yes
       5
               medium
                               Yes
           crazy loud
       6
                               Yes
       7
                 high
                                Yes
       8
                 high
                                Yes
       9
               medium
                                Yes
                                                                 comments
       0
                                                                      NaN
                                                        food tastes good
       1
       2
                                                                good food
       3
                                                My love for Korean food
```

Good food

NaN

NaN

NaN

food tasted great!

9 Steak was good, ambiance was nice and server was helpful.

	restaurant_name	target
0	NaN	like
1	Midam	like
2	pear tree	like
3	Dami	like
4	NaN	like
5	NaN	like
6	CACTUS CLUB CAFE	like
7	NaN	like
8	Frankie's	like
9	Hy's steakhouse	like

What all feature transformations you would apply on this dataset?

1.8 What did we learn today?

- Motivation to use ColumnTransformer
- ColumnTransformer syntax
- $\bullet\,$ Defining transformers with multiple transformations
- How to visualize transformed features in a dataframe
- More on ordinal features
- Different arguments OneHotEncoder
 - handle_unknow="ignore"
 - if_binary
- Dealing with text features
 - Bag of words representation: CountVectorizer

