Why generate features?

FEATURE ENGINEERING FOR MACHINE LEARNING IN PYTHON

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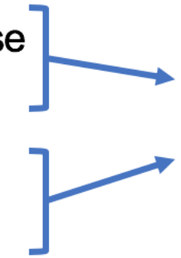
Feature Engineering

House A is a **two** bedroomed house **2000** sq. ft brownstone.

House B is **1500** sq. ft with **one** bedroom.



House	Bedrooms	sq. ft
A	2	2000
B	1	1500
...	...	...

Different types of data

- Continuous: either integers (or whole numbers) or floats (decimals)
- Categorical: one of a limited set of values, e.g. gender, country of birth
- Ordinal: ranked values, often with no detail of distance between them
- Boolean: True/False values
- Datetime: dates and times

Course structure

- Chapter 1: Feature creation and extraction
- Chapter 2: Engineering messy data
- Chapter 3: Feature normalization
- Chapter 4: Working with text features

Pandas

```
import pandas as pd
df = pd.read_csv(path_to_csv_file)
print(df.head())
```

Dataset

```
              SurveyDate  \
0    2018-02-28 20:20:00
1    2018-06-28 13:26:00
2    2018-06-06 03:37:00
3    2018-05-09 01:06:00
4    2018-04-12 22:41:00

                              FormalEducation
0    Bachelor's degree (BA. BS. B.Eng.. etc.)
1    Bachelor's degree (BA. BS. B.Eng.. etc.)
2    Bachelor's degree (BA. BS. B.Eng.. etc.)
3    Some college/university study  ...
4    Bachelor's degree (BA. BS. B.Eng.. etc.)
```

Column names

```
print(df.columns)
```

```
Index(['SurveyDate', 'FormalEducation',
       'ConvertedSalary', 'Hobby', 'Country',
       'StackOverflowJobsRecommend', 'VersionControl',
       'Age', 'Years Experience', 'Gender',
       'RawSalary'], dtype='object')
```

Column types

```
print(df.dtypes)
```

```
SurveyDate                object
FormalEducation           object
ConvertedSalary          float64
...
Years Experience           int64
Gender                    object
RawSalary                 object
dtype: object
```

Selecting specific data types

```
only_ints = df.select_dtypes(include=['int'])
print(only_ints.columns)
```

```
Index(['Age', 'Years Experience'], dtype='object')
```

Lets get going!

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Dealing with Categorical Variables

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Encoding categorical features

Index	Country
1	'India'
2	'USA'
3	'UK'
4	'UK'
5	'France'
…	…

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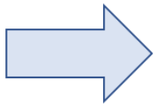

Index	C_India	C_USA	C_UK	C_France
1	1	0	0	0
2	0	1	0	0
3	0	0	1	0
4	0	0	1	0
5	0	0	0	1
…	…	…	…	…

Encoding categorical features

- One-hot encoding
- Dummy encoding

One-hot encoding

```
pd.get_dummies(df, columns=['Country'],
               prefix='C')
```

```
   C_France  C_India  C_UK  C_USA
0         0        1     0      0
1         0        0     0      1
2         0        0     1      0
3         0        0     1      0
4         1        0     0      0
```

Dummy encoding

```
pd.get_dummies(df, columns=['Country'],
               drop_first=True, prefix='C')
```

```
   C_India  C_UK  C_USA
0        1     0      0
1        0     0      1
2        0     1      0
3        0     1      0
4        0     0      0
```

One-hot vs. dummies

- **One-hot encoding**: Explainable features
- **Dummy encoding**: Necessary information without duplication

Index	Sex
0	Male
1	Female
2	Male

Index	Male	Female
0	1	0
1	0	1
2	1	0

Index	Male
0	1
1	0
2	1

Limiting your columns

```
counts = df['Country'].value_counts()
print(counts)
```

```
'USA'      8
'UK'       6
'India'    2
'France'   1
Name: Country, dtype: object
```

Limiting your columns

```
mask = df['Country'].isin(counts[counts < 5].index)

df['Country'][mask] = 'Other'

print(pd.value_counts(colors))
```

```
'USA'      8
'UK'       6
'Other'    3
Name: Country, dtype: object
```

Now you deal with categorical variables

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Numeric variables

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Types of numeric features

- Age
- Price
- Counts
- Geospatial data

Does size matter?

	Resturant_ID	Number_of_Violations
0	RS_1	0
1	RS_2	0
2	RS_3	2
3	RS_4	1
4	RS_5	0
5	RS_6	0
6	RS_7	4
7	RS_8	4
8	RS_9	1
9	RS_10	0

Binarizing numeric variables

```
df['Binary_Violation'] = 0

df.loc[df['Number_of_Violations'] > 0,
       'Binary_Violation'] = 1
```

Binarizing numeric variables

	Resturant_ID	Number_of_Violations	Binary_Violation
0	RS_1	0	0
1	RS_2	0	0
2	RS_3	2	1
3	RS_4	1	1
4	RS_5	0	0
5	RS_6	0	0
6	RS_7	4	1
7	RS_8	4	1
8	RS_9	1	1
9	RS_10	0	0

Binning numeric variables

```
import numpy as np
df['Binned_Group'] = pd.cut(
    df['Number_of_Violations'],
    bins=[-np.inf, 0, 2, np.inf],
    labels=[1, 2, 3]
)
```

Binning numeric variables

	Resturant_ID	Number_of_Violations	Binned_Group
0	RS_1	0	1
1	RS_2	0	1
2	RS_3	2	2
3	RS_4	1	2
4	RS_5	0	1
5	RS_6	0	1
6	RS_7	4	3
7	RS_8	4	3
8	RS_9	1	2
9	RS_10	0	1

Lets start practicing!

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