Feature engineering PREPROCESSING FOR MACHINE LEARNING IN PYTHON



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What is feature engineering?

Feature engineering: Creation of new features from existing ones

- Improve performance
- Insight into relationships between features
- Need to understand the data first!
- Highly dataset-dependent



Feature engineering scenarios

| ld | Text |
|----|---|
| 1 | "Feature engineering is fun!" |
| 2 | "Feature engineering is a lot of work." |
| 3 | "I don't mind feature engineering." |

| user | fav_color |
|------|-----------|
| 1 | blue |
| 2 | green |
| 3 | orange |



Feature engineering scenarios

| ld | Date |
|----|------------------|
| 4 | July 30 2011 |
| 5 | January 29 2011 |
| 6 | February 05 2011 |

| user | test1 | test2 | test3 |
|------|-------|-------|-------|
| 1 | 90.5 | 89.6 | 91.4 |
| 2 | 65.5 | 70.6 | 67.3 |
| 3 | 78.1 | 80.7 | 81.8 |



Let's practice!



Encoding categorical variables

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Categorical variables

| | user | subscribed | fav_color |
|---|------|------------|-----------|
| 0 | 1 | У | blue |
| 1 | 2 | n | green |
| 2 | 3 | n | orange |
| 3 | 4 | У | green |



Encoding binary variables - pandas

print(users["subscribed"])

print(users[["subscribed", "sub_enc"]])



| | subscribed | sub_enc |
|---|------------|---------|
| 0 | У | 1 |
| 1 | n | 0 |
| 2 | n | 0 |
| 3 | У | 1 |

users["sub_enc"] = users["subscribed"].apply(lambda val: 1 if val == "y" else 0)





Encoding binary variables - scikit-learn

from sklearn.preprocessing **import** LabelEncoder

```
le = LabelEncoder()
```

users["sub_enc_le"] = le.fit_transform(users["subscribed"])

print(users[["subscribed", "sub_enc_le"]])

| | subscribed | sub_enc_le |
|---|------------|------------|
| 0 | У | 1 |
| 1 | n | 0 |
| 2 | n | 0 |
| 3 | У | 1 |



One-hot encoding



Values: [blue, green, orange]

- blue: [1, 0, 0]
- green: [0, 1, 0]
- orange: [0, 0, 1]

| fav_color_enc |
|---------------|
| [1, 0, 0] |
| [0, 1, 0] |
| [0, 0, 1] |
| [0, 1, 0] |

< datacamp

print(users["fav_color"])

| 0 | blue |
|---|-------|
| 1 | green |

- 2 orange
- 3 green

Name: fav_color, dtype: object

print(pd.get_dummies(users["fav_color"]))

| | blue | green | orange |
|---|------|-------|--------|
| 0 | 1 | 0 | 0 |
| 1 | 0 | 1 | 0 |
| 2 | 0 | 0 | 1 |
| 3 | 0 | 1 | 0 |

R datacamp



Let's practice!



Engineering numerical features

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print(temps)

| | city | day1 | day2 | day3 |
|---|--------|------|------|------|
| 0 | NYC | 68.3 | 67.9 | 67.8 |
| 1 | SF | 75.1 | 75.5 | 74.9 |
| 2 | LA | 80.3 | 84.0 | 81.3 |
| 3 | Boston | 63.0 | 61.0 | 61.2 |

temps["mean"] = temps.loc[:,"day1":"day3"].mean(axis=1)
print(temps)

| | city | day1 | day2 | day3 | mean |
|---|--------|------|------|------|-------|
| 0 | NYC | 68.3 | 67.9 | 67.8 | 68.00 |
| 1 | SF | 75.1 | 75.5 | 74.9 | 75.17 |
| 2 | LA | 80.3 | 84.0 | 81.3 | 81.87 |
| 3 | Boston | 63.0 | 61.0 | 61.2 | 61.73 |

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Dates

print(purchases)

| | | | date | purchase |
|---|----------|----|------|----------|
| 0 | July | 30 | 2011 | \$45.08 |
| 1 | February | 01 | 2011 | \$19.48 |
| 2 | January | 29 | 2011 | \$76.09 |
| 3 | March | 31 | 2012 | \$32.61 |
| 4 | February | 05 | 2011 | \$75.98 |





Dates

purchases["date_converted"] = pd.to_datetime(purchases["date"]) purchases['month'] = purchases["date_converted"].dt.month print(purchases)

| | | | date | purchase | date_converted | month |
|---|----------|----|------|----------|----------------|-------|
| 0 | July | 30 | 2011 | \$45.08 | 2011-07-30 | 7 |
| 1 | February | 01 | 2011 | \$19.48 | 2011-02-01 | 2 |
| 2 | January | 29 | 2011 | \$76.09 | 2011-01-29 | 1 |
| 3 | March | 31 | 2012 | \$32.61 | 2012-03-31 | 3 |
| 4 | February | 05 | 2011 | \$75.98 | 2011-02-05 | 2 |





Let's practice!



Engineering text features

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Extraction

• **Regular expressions**: code to identify patterns

import re

```
my_string = "temperature:75.6 F"
```

```
temp = re.search("\d+\.\d+", my_string)
```

```
print(float(temp.group(0)))
```

75.6





Vectorizing text

TF/IDF: Vectorizes words based upon importance

- TF = Term Frequency
- IDF = Inverse Document Frequency



Vectorizing text

from sklearn.feature_extraction.text import TfidfVectorizer

print(documents.head())

| 0 | Building on successful events last summer and |
|---|--|
| 1 | Build a website for an Afghan business |
| 2 | Please join us and the students from Mott Hall |
| 3 | The Oxfam Action Corps is a group of dedicated |
| 4 | Stop 'N' Swap reduces NYC's waste by finding n |

```
tfidf_vec = TfidfVectorizer()
text_tfidf = tfidf_vec.fit_transform(documents)
```





Text classification

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$



Let's practice!

