원핫 인코딩 실습

학습 목표

01 pd.get_dummy를 이용한 원핫 인코딩 실습02 성인 인구조사 소득 데이터 셋(adult.data)을 활용한 onehot encoding실습

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- 01. 원핫 인코딩 실습
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01. 원핫 인코딩 실습

목차로 이동하기

- 데이터 셋을 불러와 원핫 인코딩 실습
- hello world 원핫 인코딩 실습

```
In [1]: import mglearn
import pandas as pd
import os
```

In [2]: demo_df = pd.DataFrame({"Product":['양말', '여우', '양말', '상자']}) display(demo_df)

Product 0 양말 1 여우 2 양말 3 상자

```
In [3]: onehot = pd.get_dummies(demo_df)
  onehot
```

 Out[3]:
 Product_상자
 Product_양말
 Product_여우

 0
 0
 1
 0

 1
 0
 0
 1

 2
 0
 1
 0

 3
 1
 0
 0

```
In [4]: df = pd.concat([demo_df, onehot], axis=1)
    df
```

t[4]:		Product	Product_상자	Product_양말	Product_여우
	0	양말	0	1	0
	1	여우	0	0	1
	2	양말	0	1	0
	3	상자	1	0	0

02. adult.data 셋을 활용한 onehot encoding 실습

목차로 이동하기

Ou

```
In [5]: path = os.path.join(mglearn.datasets.DATA_PATH, 'adult.data')
         print(path)
        C:\Users\totofriend\anaconda3\lib\site-packages\mglearn\data\adult.data
In [6]: data = pd.read_csv(path,
                        header=None,
                        index_col=False,
                        names=['age', 'workclass', 'fnlwgt', 'education',
                               'education-num', 'marital-status', 'occupation', 'relationship
'race', 'gender', 'capital-gain', 'capital-loss',
'hours-per-week', 'native-country', 'income'])
         data.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 32561 entries, 0 to 32560
        Data columns (total 15 columns):
             Column
                             Non-Null Count
                                              Dtype
         0
             age
                              32561 non-null
                                              int64
                              32561 non-null
                                              object
         1
             workclass
         2
             fnlwgt
                              32561 non-null
                                              int64
             education
                              32561 non-null
                                              object
             education-num
                              32561 non-null
                                              int64
             marital-status 32561 non-null
                                              object
                              32561 non-null
             occupation
                                              object
             relationship
                              32561 non-null
                                              object
                              32561 non-null
             race
                                              object
         9
                              32561 non-null
                                              object
              gender
          10 capital-gain
                              32561 non-null
                                              int64
             capital-loss
                              32561 non-null
                                              int64
          12 hours-per-week 32561 non-null
                                              int64
          13 native-country 32561 non-null
                                              object
                              32561 non-null
          14 income
                                              object
         dtypes: int64(6), object(9)
        memory usage: 3.7+ MB
        data.columns
In [8]:
        Out[8]:
                'capital-gain', 'capital-loss', 'hours-per-week', 'native-country',
                'income'],
               dtype='object')
```

일부 변수 선택 후, 진행

Out[9]:		age	workclass	education	gender	hours-per-week	occupation	income
	0	39	State-gov	Bachelors	Male	40	Adm-clerical	<=50K
	1	50	Self-emp-not-inc	Bachelors	Male	13	Exec-managerial	<=50K
	2	38	Private	HS-grad	Male	40	Handlers-cleaners	<=50K
	3	53	Private	11th	Male	40	Handlers-cleaners	<=50K
	4	28	Private	Bachelors	Female	40	Prof-specialty	<=50K

의미 있는 범주형 데이터 있는지 확인

```
In [10]: print(data.gender.value_counts())
```

Male 21790 Female 10771

Name: gender, dtype: int64

pandas에서 get_dummies 함수를 이용하여 인코딩

```
In [11]: print("원본 특성 :₩n", list(data.columns), "₩n")
          data_dummies = pd.get_dummies(data)
          print("get_dummies 후 특성 : ₩n", list(data_dummies.columns))
          원본 특성 :
           ['age', 'workclass', 'education', 'gender', 'hours-per-week', 'occupation', 'incom
          e'l
          get_dummies 후 특성 :
          ['age', 'hours-per-week', 'workclass_?', 'workclass_ Federal-gov', 'workclass_ Loc
          al-gov', 'workclass_ Never-worked', 'workclass_ Private', 'workclass_ Self-emp-inc',
          'workclass_ Self-emp-not-inc', 'workclass_ State-gov', 'workclass_ Without-pay', 'ed
          ucation_ 10th', 'education_ 11th', 'education_ 12th', 'education_ 1st-4th', 'education_ 5th-6th', 'education_ 7th-8th', 'education_ 9th', 'education_ Assoc-acdm', 'educ
          ation_ Assoc-voc', 'education_ Bachelors', 'education_ Doctorate', 'education_ HS-gr
          ad', 'education_ Masters', 'education_ Preschool', 'education_ Prof-school', 'educat
          ion_ Some-college', 'gender_ Female', 'gender_ Male', 'occupation_ ?', 'occupation_
          Adm-clerical', 'occupation_ Armed-Forces', 'occupation_ Craft-repair', 'occupation_
          Exec-managerial', 'occupation_ Farming-fishing', 'occupation_ Handlers-cleaners', 'o
          ccupation_ Machine-op-inspct', 'occupation_ Other-service', 'occupation_ Priv-house-
          serv', 'occupation_ Prof-specialty', 'occupation_ Protective-serv', 'occupation_ Sal
          es', 'occupation_ Tech-support', 'occupation_ Transport-moving', 'income_ <=50K', 'i
          ncome_ >50K']
```

• age와 hours-per-week는 그대로이지만 범주형 특성은 새로운 특성으로 확장

특성을 포함한 열 'age'~'occupation_ Transport-moving' 모두 추출

```
In [12]: features = data_dummies.loc[:, "age":"occupation_ Transport-moving"]
X = features.values
y = data_dummies['income_ >50K'].values
```

```
In [13]: print("X.shape : {}, y.shape : {}".format(X.shape, y.shape))

X.shape : (32561, 44), y.shape : (32561,)
```

실습 1

- 로지스틱 모델을 만들어보기
 - (1) 데이터를 나누어준다.
 - (2) 모델을 만든다.
 - (3) 모델을 학습한다.(학습 데이터를 이용해서)
 - (4) score를 확인(테스트 데이터를 이용해서)

로지스틱 모델 사용해 보기

```
In [14]: from sklearn.linear_model import LogisticRegression
         from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=0)
         logreg = LogisticRegression()
         logreg.fit(X_train, y_train)
         C:\Users\totofriend\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:81
         4: ConvergenceWarning: lbfgs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as shown in:
             https://scikit-learn.org/stable/modules/preprocessing.html
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
           n_iter_i = _check_optimize_result(
        LogisticRegression()
Out[14]:
In [15]: print("학습용 점수 {:.2f}".format(logreg.score(X_train, y_train)))
         print("테스트 점수 {:.2f}".format(logreg.score(X_test, y_test)))
         학습용 점수 0.81
         테스트 점수 0.81
In [16]: from sklearn.ensemble import RandomForestClassifier
         from sklearn.neighbors import KNeighborsClassifier
```

랜덤 포레스트를 활용한 모델 구축

```
In [19]: model = RandomForestClassifier().fit(X_train, y_train)

print("학습용 점수 {:.2f}".format(model.score(X_train, y_train)))
print("테스트 점수 {:.2f}".format(model.score(X_test, y_test)))

학습용 점수 0.94
테스트 점수 0.79
```

knn모델을 활용한 모델 구축

```
In [18]: model = KNeighborsClassifier().fit(X_train, y_train)

print("학습용 점수 {:.2f}".format(model.score(X_train, y_train)))
print("테스트 점수 {:.2f}".format(model.score(X_test, y_test)))
```

학습용 점수 0.84 테스트 점수 0.78