ch02 지도학습 - knn - 실습

- Machine Learning with sklearn @ DJ,Lim
- 최종 update : 22/05

학습 내용

- titanic 데이터 셋을 활용하여 knn 모델을 구현한다.
- 가장 높은 일반화 성능을 갖는 k의 값은 무엇인지 찾아보자.

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01 데이터 준비

```
In [ ]: import pandas as pd
    from sklearn.model_selection import train_test_split
    import numpy as np
```

```
In [ ]: dat = pd.read_csv("train.csv")
    dat
```

Out[]:	P	assengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Far
	0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.250
	1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.283:
	2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.925(
	3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.100(
	4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.050(
	•••										• (
	886	887	0	2	Montvila, Rev. Juozas	male	27.0	0	0	211536	13.0000

Far	Ticket	Parch	SibSp	Age	Sex	Name	Pclass	Survived	PassengerId	
30.000	112053	0	0	19.0	female	Graham, Miss. Margaret Edith	1	1	888	887
23.450	W./C. 6607	2	1	NaN	female	Johnston, Miss. Catherine Helen "Carrie"	3	0	889	888
30.000	111369	0	0	26.0	male	Behr, Mr. Karl Howell	1	1	890	889
7.750	370376	0	0	32.0	male	Dooley, Mr. Patrick	3	0	891	890

891 rows × 12 columns

```
# 컬럼(or feature)명 확인
In [ ]:
         dat.columns
Out[ ]: Index(['PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibSp', 'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'],
               dtype='object')
         dat.info()
In [ ]:
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 891 entries, 0 to 890
         Data columns (total 12 columns):
          #
              Column
                            Non-Null Count Dtype
                            -----
          0
              PassengerId 891 non-null
                                             int64
          1
              Survived
                            891 non-null
                                             int64
          2
              Pclass
                            891 non-null
                                             int64
          3
              Name
                            891 non-null
                                             object
          4
              Sex
                            891 non-null
                                             object
          5
              Age
                            714 non-null
                                             float64
          6
              SibSp
                            891 non-null
                                             int64
          7
              Parch
                            891 non-null
                                             int64
          8
              Ticket
                            891 non-null
                                             object
          9
              Fare
                            891 non-null
                                             float64
          10 Cabin
                            204 non-null
                                             object
          11 Embarked
                            889 non-null
                                             object
         dtypes: float64(2), int64(5), object(5)
```

데이터 선택 및 나누기

memory usage: 83.7+ KB

- test_size : 테스트 데이터 셋 비율 선택
- random_state : 데이터을 뽑을 때, 지정된 패턴으로 선택

```
In []: X = dat[['Pclass', 'SibSp']]
y = dat['Survived']

# 90%: 학습용, 10%: 테스트용
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.1,
```

```
random state=0)
        X_train.shape, X_test.shape, y_train.shape, y_test.shape
Out[]: ((801, 2), (90, 2), (801,), (90,))
        from sklearn.neighbors import KNeighborsClassifier
In [ ]:
        model = KNeighborsClassifier(n neighbors=2)
        model.fit(X train, y train)
        ### 예측시키기
        pred = model.predict(X test)
        (pred == y test).sum() / len(pred)
In [ ]:
Out[]: 0.6
In []: print("테스트 세트의 정확도 : {:.2f}".format(np.mean(pred == y test)))
        테스트 세트의 정확도 : 0.60
       결측치 처리 및 레이블 인코딩
In [ ]: | dat.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 891 entries, 0 to 890
        Data columns (total 13 columns):
                        Non-Null Count Dtype
        #
            Column
                         -----
            ----
        ___
            PassengerId 891 non-null
         0
                                        int64
            Survived
         1
                        891 non-null
                                       int64
         2
            Pclass
                         891 non-null
                                       int64
                                     object
         3
            Name
                        891 non-null
```

dtypes: float64(2), int64(6), object(5) memory usage: 90.6+ KB In []: dat.head()

object

int64

int64

object

float64

object

object

int64

float64

891 non-null

891 non-null

891 non-null

891 non-null

891 non-null

891 non-null

204 non-null

891 non-null

889 non-null

4

5

6

7

8

9

Sex

Age

SibSp

Parch

Fare

11 Embarked 12 Sex num

10 Cabin

Ticket

Out[]:		PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
	0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500
	1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833
	2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500

map함수

- [Series].map(함수 또는 변경값): Series를 대상으로 원하는 함수 적용 또는 값을 대체
- 값으로 dict, Series를 대상으로 한다.
- https://pandas.pydata.org/docs/reference/api/pandas.Series.map.html

```
In [ ]: mapping = { "male":1, 'female':2 }
    dat['Sex_num'] = dat['Sex'].map(mapping)
    dat.head()
```

Out[]:		PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
	0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500
	1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833
	2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250
	3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000
	4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500

```
In []: # [].fillna() : 결측값을 채운다.
val_mean = dat['Age'].mean()
dat['Age'] = dat['Age'].fillna( val_mean )
dat.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 13 columns):
# Column Non-Null Count Dtype
```

```
0  PassengerId 891 non-null int64
1  Survived 891 non-null int64
2  Pclass 891 non-null int64
3  Name 891 non-null object
4  Sex 891 non-null object
5  Age 891 non-null float64
6  SibSp 891 non-null int64
7  Parch 891 non-null int64
8  Ticket 891 non-null object
9  Fare 891 non-null float64
10  Cabin 204 non-null object
11  Embarked 889 non-null object
12  Sex_num 891 non-null int64
dtypes: float64(2), int64(6), object(5)
memory usage: 90.6+ KB
```

학습, 테스트 데이터 셋 나누기

```
In []: X = dat[['Pclass', 'SibSp', 'Sex_num', 'Age']]
y = dat['Survived']

# 90%: 학습용, 10%: 테스트용
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.1, random_state=0)

X_train.shape, X_test.shape, y_train.shape, y_test.shape

Out[]: ((801, 4), (90, 4), (801,), (90,))

In []: from sklearn.neighbors import KNeighborsClassifier model = KNeighborsClassifier(n_neighbors=2) model.fit(X_train, y_train)
### 예측시키기
pred = model.predict(X_test)
print("테스트 세트의 정확도: {:.2f}".format(np.mean(pred == y_test)))
```

k값에 따른 정확도 확인

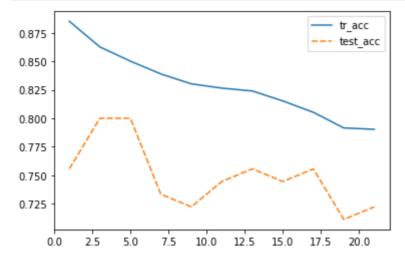
테스트 세트의 정확도 : 0.76

```
In [ ]: | tr_acc = []
        test acc = []
        k nums = range(1, 22, 2)# 1,3,5\sim21
         for n in k_nums:
             # 모델 선택 및 학습
             model = KNeighborsClassifier(n neighbors=n)
             model.fit(X_train, y_train)
             # 정확도 구하기
             acc tr = model.score(X train, y train)
             acc test = model.score(X test, y test)
             # 정확도 값 저장。
             tr acc.append(acc tr)
             test_acc.append(acc_test)
             print("k : ", n)
             print("학습용셋 정확도 {:.3f}".format(acc tr) )
             print("테스트용셋 정확도 {:.3f}".format(acc test) )
```

k : 1 학습용셋 정확도 0.885

```
테스트용셋 정확도 0.756
       k: 3
        학습용셋 정확도 0.863
       테스트용셋 정확도 0.800
       k : 5
       학습용셋 정확도 0.850
       테스트용셋 정확도 0.800
       k : 7
        학습용셋 정확도 0.839
       테스트용셋 정확도 0.733
       k: 9
        학습용셋 정확도 0.830
       테스트용셋 정확도 0.722
       k: 11
       학습용셋 정확도 0.826
       테스트용셋 정확도 0.744
       k : 13
        학습용셋 정확도 0.824
       테스트용셋 정확도 0.756
       k: 15
       학습용셋 정확도 0.815
       테스트용셋 정확도 0.744
       k: 17
        학습용셋 정확도 0.805
       테스트용셋 정확도 0.756
       k: 19
       학습용셋 정확도 0.792
       테스트용셋 정확도 0.711
       k : 21
       학습용셋 정확도 0.790
       테스트용셋 정확도 0.722
In [ ]: | import seaborn as sns
        print(sns. version )
       0.11.0
In [ ]: | # tr_acc = []
        # test acc = []
        dat = { "tr_acc":tr_acc, "test_acc":test_acc }
        data df = pd.DataFrame(dat, index=range(1, 22, 2))
        data df
Out[]: tr_acc test_acc
         1 0.885144 0.755556
        3 0.862672 0.800000
        5 0.850187 0.800000
        7 0.838951 0.733333
        9 0.830212 0.722222
        11 0.826467 0.744444
        13 0.823970 0.755556
        15 0.815231 0.744444
        17 0.805243 0.755556
          0.791511 0.711111
        19
        21 0.790262 0.722222
In [ ]: | import matplotlib.pyplot as plt
```

```
In [ ]: sns.lineplot(data=data_df, palette="tab10")
   plt.show()
```



REF

• map 메소드 : https://pandas.pydata.org/docs/reference/api/pandas.Series.map.html