# 이번 수업에서는

- 1. 행/ 열 추가
- 2. 원소 선택
- 3. 원소 변경

## 행 추가

객체.loc["새로운 행의 이름"] = 데이터값들 (리스트 또는 배열)



	수학	영어	음악	체육
서준	90	98	85	100
우현	80	89	95	90
인아	70	95	100	90
상기	95	100	80	95

## 열 추가

→ 객체["새로운 열의 이름"] = 데이터값들 (리스트 또는 배열)

df["미술"] = [80, 90, 95, 100]

	수학	영 어	음악	체육	미술
서준	90	98	85	100	80
우현	80	89	95	90	90
인아	70	95	100	90	95
상기	95	100	80	95	100

#### 열 추가

→ 객체["새로운 열의 이름"] = 데이터값들 (리스트 또는 배열)

ValueError: Length of values does not match length of index

	수학	영어	음악	체육	미술	과학		
서준	90	98	85	100	80	80		
우현	80	89	95	90	90	80	/	
인아	70	95	100	90	95	80		
상기	95	100	80	95	100	80		

#### 원소 선택(조회)

객체.iloc[ 행번호, 열번호 ]

df[2,3] # pandas는 numpy 가 아니다 df.iloc[2,3]

#### KeyError

df.iloc[2][3]

/usr/local/lib/python3.6/dist-package

2896 try:

-> 2897 return self.\_ 2898 except KeyError:

df.iloc[2,3] df.iloc[2][3]

90 90

#### 원소 선택

▶ 객체.loc[ "행이름", "열이름" ]

```
df.loc["인아", "체육"]
```

```
df.loc["인아", "체육"]
```

90

```
df.loc["인아", ["체육", "영어"]] # df.iloc[2, [3, 1]]
```

```
df.loc["인아", ["체육", "영어"] ]
```

체육 90 영어 95

Name: 인아, dtype: int64

#### 원소 값 바꾸기

▶ 원소 선택 = 새로운 값

# 인아의 체육점수를 95점으로 df.loc["인아", "체육"] = 95

df.iloc[2, 3] = 90

df

## 원소 값 바꾸기

Extract a diagonal or construct a diagona

여러 개를 바꾸려면

df.loc["인아", ["체육", "영어"]] = 80, 90

	수학	영어	음악	체육	미술
서준	90	98	85	100	80
우현	80	89	95	90	90
인아	70	90	100	80	95
상기	95	100	80	95	100

# 이번 수업에서는

1. 파일 읽기. csv, Excel 쓰기

2. 데이터 프레임 살펴보기

3. 기본 통계함수 적용해 보기

## csv 파일 읽기

df = pd.read\_csv("https://raw.githubusercontent.com/plotly/datasets/master/auto-mpg.csv")
df

	mpg	cylinders	displacement	horsepower	weight	acceleration	model-year
0	18.0	8	307.0	130.0	3504	12.0	70
1	15.0	8	350.0	165.0	3693	11.5	70
2	18.0	8	318.0	150.0	3436	11.0	70
3	16.0	8	304.0	150.0	3433	12.0	70

df.to\_csv("mpg.csv")

#### excel 파일 읽기

df = pd.read\_excel('http://qrc.depaul.edu/Excel\_Files/Presidents.xls')
df

	President	Years in office	Year first inaugurated	Age at inauguration	State elected from	# of electoral votes
0	George Washington	8.0	1789	57	Virginia	69
1	John Adams	4.0	1797	61	Massachusetts	132
2	Thomas Jefferson	8.0	1801	57	Virginia	73

df.to\_excel("President.xlsx")

- head( )
- tail( )
- •/describe( )

- / shape
  - dtypes

df = pd.read\_csv("https://raw.githubusercontent.com/plotly/datasets/master/auto-mpg.csv")
df

	mpg	cylinders	displacement	horsepower	weight	acceleration	model-year
0	18.0	8	307.0	130.0	3504	12.0	70
1	15.0	8	350.0	165.0	3693	11.5	70
2	18.0	8	318.0	150.0	3436	11.0	70
3	16.0	8	304.0	150.0	3433	12.0	70

#### df.head()

	mpg	cylinders	displacement	horsepower	weight	acceleration	model-year
0	18.0	8	307.0	130.0	3504	12.0	70
1	15.0	8	350.0	165.0	3693	11.5	70
2	18.0	8	318.0	150.0	3436	11.0	70
3	16.0	8	304.0	150.0	3433	12.0	70
4	17.0	8	302.0	140.0	3449	10.5	70

#### df.tail()

	mpg	cylinders	displacement	horsepower	weight	acceleration	model-year
393	27.0	4	140.0	86.0	2790	15.6	82
394	44.0	4	97.0	52.0	2130	24.6	82
395	32.0	4	135.0	84.0	2295	11.6	82
396	28.0	4	120.0	79.0	2625	18.6	82
397	31.0	4	119.0	82.0	2720	19.4	82

#### df.tai1(2)

	mpg	cylinders	displacement	horsepower	weight	acceleration	model-year
396	28.0	4	120.0	79.0	2625	18.6	82
397	31.0	4	119.0	82.0	2720	19.4	82

#### df.info()

df.shape

(398, 7)

#### df.dtypes

mpg	float64
cylinders	int64
displacement	float64
horsepower	float64
weight	int64
acceleration	float64
model-year	int64
dtype: object	

df.size

2786

## ➤ describe() 기초통계량

df.describe()

	mpg	cylinders	displacement	horsepower	weight	acceleration
count	398.000000	398.000000	398.000000	396.000000	398.000000	398.000000
mean	23.514573	5.454774	193.425879	104.189394	2970.424623	15.568090
std	7.815984	1.701004	104.269838	38.402030	846.841774	2.757689
min	9.000000	3.000000	68.000000	46.000000	1613.000000	8.000000
25%	17.500000	4.000000	104.250000	75.000000	2223.750000	13.825000
50%	23.000000	4.000000	148.500000	92.000000	2803.500000	15.500000
75%	29.000000	8.000000	262.000000	125.000000	3608.000000	17.175000
max	46.600000	8.000000	455.000000	230.000000	5140.000000	24.800000

➤ count( ) 빈도수

value\_counts() 중복제거한 unique 한 개수

#### df.count()

mpg	398
cylinders	398
displacement	398
horsepower	396
weight	398
acceleration	398
model-year	398
dtype: int64	

df['cylinders'].value\_counts()

```
4 204
8 103
6 84
3 4
5 3
```

Name: cylinders, dtype: int64

## 기초통계량 직접 계산하기

d <b>£</b> ====()		df.std()		df.median()	
df.mean()		mpg	7.8159		23.0
mpg cylinders displacement horsepower weight acceleration model-year dtype: float64	23.514573 5.454774 193.425879 104.189394 2970.424623 15.568090 76.010050	cylinders displacement horsepower weight acceleration model-year dtype: float64	1.7010 104.2698 38.4020 846.8417 2.7570 3.6970	displacement horsepower weight acceleration	4.0 148.5 92.0 2803.5 15.5 76.0
derland 11 magnet		df.var()		df.max()	
df['mpg'].mean(	J	mpg	61.0	mpg	46.6
23.514572864321	.615	cylinders displacement horsepower weight acceleration	2.8 10872.1 1474.1 717140.9	cylinders displacement horsepower	8.0 455.0 230.0 5140.0 24.8

model-year

dtype: float64

13. model-year

dtype: float64

82.0

# 상관계수 구하기

## df.corr()

	mpg	cylinders	displacement	horsepower	weight	acceleration	model- year
mpg	1.000000	-0.775396	-0.804203	-0.777575	-0.831741	0.420289	0.579267
cylinders	-0.775396	1.000000	0.950721	0.843751	0.896017	-0.505419	-0.348746
displacement	-0.804203	0.950721	1.000000	0.897787	0.932824	-0.543684	-0.370164
horsepower	-0.777575	0.843751	0.897787	1.000000	0.864350	-0.687241	-0.420697
weight	-0.831741	0.896017	0.932824	0.864350	1.000000	-0.417457	-0.306564
acceleration	0.420289	-0.505419	-0.543684	-0.687241	-0.417457	1.000000	0.288137
model-year	0.579267	-0.348746	-0.370164	-0.420697	-0.306564	0.288137	1.000000

## 상관계수 구하기

df[["mpg", "cylinders", "displacement"]].corr()

	mpg	cylinders	displacement
mpg	1.000000	-0.775396	-0.804203
cylinders	-0.775396	1.000000	0.950721
displacement	-0.804203	0.950721	1.000000

# 이번 수업에서는

결측치

### 결측치란

## • 결측치란

```
import seaborn as sns
import pandas as pd
import numpy as np

df = sns.load_dataset('titanic')
df.head()
```

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male
0	0	3	male	22.0	1	0	7.2500	S	Third	man	True
1	1	1	female	38.0	1	0	71.2833	С	First	woman	False
2	1	3	female	26.0	0	0	7.9250	S	Third	woman	False
3	1	1	female	35.0	1	0	53.1000	S	First	woman	False
4	0	3	male	35.0	0	0	8.0500	S	Third	man	True

#### df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 15 columns):
survived
              891 non-null int64
              891 non-null int64
pclass
              891 non-null object
sex
              714 non-null float64
age
              891 non-null int64
sibsp
parch
              891 non-null int64
fare
              891 non-null float64
embarked
              889 non-null object
class
              891 non-null category
              891 non-null object
who
              891 non-null bool
adult_male
deck
              203 non-null category
embark_town
              889 non-null object
alive
              891 non-null object
alone
              891 non-null bool
dtypes: bool(2), category(2), float64(2), int64(4), object(5)
memory usage: 80.6+ KB
```

# 결측치 존재

# df.isnull()

	survived	pclass	sex	age	S
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	

df.isnull().sum()

survived	0
pclass	0
sex	0
age	177
sibsp	0
parch	0
fare	0
embarked	2
class	0
who	0
adult_male	0
deck	688
embark_town	2
alive	0
alone	0
dtype: int64	

#### 결측치 삭제

df1 =df.copy()

#### df1.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 891 entries, 0 to 890 Data columns (total 15 columns): 891 non-null int64 survived 891 nen\_null int64 pclass 891 non-null object sex 714 non-null float64 age 891 non-null int64 sibsp parch 891 non-null int64 fare 891 non-null float64 embarked 889 non-null object class 891 non-null category who 891 non-null object adult\_male 891 non-null bool deck 203 non-null category



df1.dropna().info() # 행 삭제

<class 'pandas.core.frame.DataFrame'> Int64Index: 182 entries, 1 to 889 Data columns (total 15 columns): survived 182 non-null int64 182 non-null int64 pclass 182 non-null object sex 182 non-null float64 age sibsp 182 non-null int64 182 non-null int64 parch fare 182 non-null float64 embarked 182 non-null object class 182 non-null category 182 non-null object who adult\_male 182 non-null bool deck 182 non-null category

#### 결측치 삭제

```
df1.dropna(axis =1).info() # 열 식제
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 891 entries, 0 to 890
Data columns (total 11 columns):
             891 non-null int64
survived
pclass
             891 non-null int64
             891 non-null object
sex
             891 non-null int64
sibsp
             891 non-null int64
parch
fare
             891 non-null float64
class
             891 non-null category
who
             891 non-null object
             891 non-null bool
adult male
alive
             891 non-null object
alone
             891 non-null bool
dtypes: bool(2), category(1), float64(1), int64(4), object(3)
memory usage: 58.5+ KB
```

#### 결측치 대체

```
# 평균값으로 대체하기
mean_age = df2['age'].mean()
df2['age'].fillna(mean_age, inplace= True)
df2['age'].isnull()
      False
0
      False
     False
      False
      False
886
      False
887
    False
888 False
889 False
890
    False
                                         df2['age'].isnull().sum()
Name: age, Length: 891, dtype: bool
```

### 중복데이터

A B C

a 1 1

a 1 1

b 1 2

a 2 2

b 2 2

### df.duplicated()

```
A B C
0 a 1 1
1 a 1 1
2 b 1 2
3 a 2 2
4 b 2 2
```

• 중복된 행의 개수

```
df.duplicated()

0 False
1 True
2 False
3 False
4 False
dtype: bool
```

```
] df['A'].duplicated() # 한개 열(vector)에도 적용
```

```
0 False
1 True
2 False
3 True
4 True
Name: A, dtype: bool
```

## 중복행 제거

A B C

a 1 1

a 1 1

b 1 2

a 2 2

b 2 2

df.drop\_duplicates()

А В С

a 1 1

b 1 2

a 2 2

b 2 2

• 저장하려면 inplace = True

# 이번 수업에서는

데이터프레임 합치기 (concat)

#### concat

Series + Series

```
E = pd.Series(['e0','e1','e2','e3'], name = 'e')
F = pd.Series(['f0','f1','f2'], name = 'f', index = [3,4,5])
G = pd.Series(['g0','g1','g2','g3'], name = 'g')
```

E		F	G
0 1 2	e0 e1 e2	3 f0 4 f1 5 f2	0 g0 1 g1 2 g2
3	e3	Name: f,	2 g2 3 g3
Nam	e: e,		Name: g,

• axis = 0 이 기본값이다

```
pd.concat([E, G])
pd.concat([E, F])
     e0
                     0
                           e0
                           e1
     e1
                          e2
   e2
                     3
    e3
                          e3
   f0
                          g0
                     0
                          g1
    f1
5
     f2
                     2
                          g2
                     3
                          g3
```

• index에 맞게 병합

pd.concat([E, G], axis = 1)

e g

**0** e0 g0

**1** e1 g1

**2** e2 g2

**3** e3 g3

pd.concat([E, F], axis = 1)

e f

**0** e0 NaN

1 e1 NaN

2 e2 NaN

**3** e3 f0

4 NaN f1

**5** NaN f2

• 결과물은 Series 이거나 DataFrame 이다

```
type(pd.concat([E, G], axis = 0 ) )
pandas.core.series.Series
```

```
type(pd.concat([E, G], axis = 1))
```

pandas.core.frame.DataFrame

#### concat

DataFrame + DataFrame

```
df1
```

```
a b c
```

b c d

## pd.concat([df1,df2])

a b c d

**0** a0 b0 c0 NaN

**1** a1 b1 c1 NaN

2 a2 b2 c2 NaN

**1** NaN b2 c2 d2

2 NaN b3 c3 d3

**3** NaN b4 c4 d4

df1

a b c

**0** a0 b0 c0

**1** a1 b1 c1

**2** a2 b2 c2

df2

b c d

**1** b2 c2 d2

**2** b3 c3 d3

**3** b4 c4 d4

```
pd.concat([df1,df2], ignore_index=True)
```

- a b c d
- a0 b0 c0 NaN
- a1 b1 c1 NaN
- a2 b2 c2 NaN
- NaN b2 c2 d2
- 4 NaN b3 c3 d3
- NaN b4 c4 d4

pd.concat([df1,df2], axis = 1)

b c b c d a b0 c0 NaN a0 NaN NaN 0 1 b2 a1 b1 c1 c2 d2 b2 c2 b3 c3 2 a2 d3 d4 c4 b4 3 NaN NaN NaN

df1

a b c

**0** a0 b0 c0

**1** a1 b1 c1

**2** a2 b2 c2

df2

b c d

**1** b2 c2 d2

**2** b3 c3 d3

**3** b4 c4 d4

pd.concat([df1,df2], axis = 1, join = 'inner')

a b c b c d

- **1** a1 b1 c1 b2 c2 d2
- 2 a2 b2 c2 b3 c3 d3