#2018107115 고지영

# 인공지능 발표



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#1, 주제

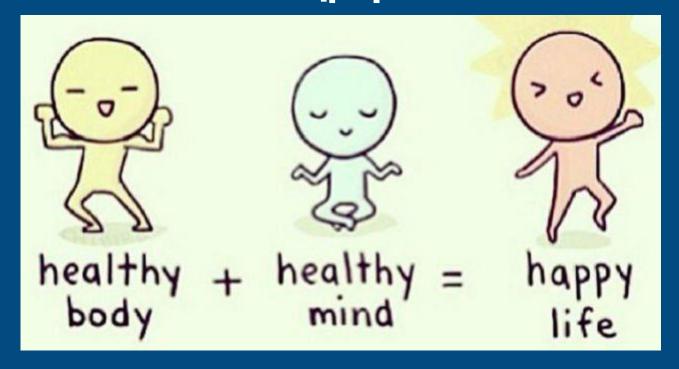
#2, 데이터 읽어오기

#3, 데이터 표시

#4, 학습 및 테스트



# 여러 요인에 따라 달라지는 치료비용 예측



### #1 사용한 데이터



#### Context

Machine Learning with R by Brett Lantz is a book that provides an introduction to machine learning using R. As far as I can tell, Packt Publishing does not make its datasets available online unless you buy the book and create a user account which can be a problem if you are checking the book out from the library or borrowing the book from a friend. All of these datasets are in the public domain but simply needed some cleaning up and recoding to match the format in the book.

#### Content

#### Columns

- · age: age of primary beneficiary
- · sex: insurance contractor gender, female, male
- bmi: Body mass index, providing an understanding of body, weights that are relatively high or low relative to height, objective index of body weight (kg / m ^ 2) using the ratio of height to weight, ideally 18.5 to 24.9
- · children: Number of children covered by health insurance / Number of dependents
- · smoker: Smoking
- · region: the beneficiary's residential area in the US, northeast, southeast, southwest, northwest.
- · charges: Individual medical costs billed by health insurance

#### Acknowledgements

The dataset is available on GitHub here.

#### Inspiration

Can you accurately predict insurance costs?

#### 의료비 개인 데이터 세트 (선형회귀를 통한 보험료 예측)

#### 컬럼

- 나이
- 성별(남,여)
- Bmi(체질량 지수 정상:18.5~24.9)
- 자녀(자녀의 수)
- 지역(미국)
- 흡연(흡연자,비흡연자)
- 비용(개인 의료비)

#### #2 데이터 읽어오기

```
import numpy as np
import pandas as pd
import os
import matplotlib.pyplot as pl
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
data = pd.read_csv('../input/insurance.csv')
```

모듈 import 데이터 읽어 오기

age         sex         bmi         children         smoker         region         charges           0         19         female         27.900         0         yes         southwest         16884.92400           1         18         male         33.770         1         no         southeast         1725.55230           2         28         male         33.000         3         no         southeast         4449.46200           3         33         male         22.705         0         no         northwest         21984.47061           4         32         male         28.880         0         no         northwest         3756.62160           5         31         female         25.740         0         no         southeast         3756.62160           6         46         female         33.440         1         no         southeast         8240.58960           7         37         female         27.740         3         no         northwest         7281.50560           8         37         male         29.830         2         no         northwest         6406.41070           9         60         female <th colspan="9">data.head(10)</th>	data.head(10)								
1         18         male         33.770         1         no         southeast         1725.55230           2         28         male         33.000         3         no         southeast         4449.46200           3         33         male         22.705         0         no         northwest         21984.47061           4         32         male         28.880         0         no         northwest         3866.85520           5         31         female         25.740         0         no         southeast         3756.62160           6         46         female         33.440         1         no         southeast         8240.58960           7         37         female         27.740         3         no         northwest         7281.50560           8         37         male         29.830         2         no         northwest         6406.41070		age	sex	bmi	children	smoker	region	charges	
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3 33 male 22.705 0 no northwest 21984.47061 4 32 male 28.880 0 no northwest 3866.85520 5 31 female 25.740 0 no southeast 3756.62160 6 46 female 33.440 1 no southeast 8240.58960 7 37 female 27.740 3 no northwest 7281.50560 8 37 male 29.830 2 no northeast 6406.41070	1	18	male	33.770	1	no	southeast	1725.55230	
4 32 male 28.880 0 no northwest 3866.85520 5 31 female 25.740 0 no southeast 3756.62160 6 46 female 33.440 1 no southeast 8240.58960 7 37 female 27.740 3 no northwest 7281.50560 8 37 male 29.830 2 no northeast 6406.41070	2	28	male	33.000	3	no	southeast	4449.46200	
5     31 female     25.740     0     no     southeast     3756.62160       6     46 female     33.440     1     no     southeast     8240.58960       7     37 female     27.740     3     no     northwest     7281.50560       8     37 male     29.830     2     no     northeast     6406.41070	3	33	male	22.705	0	no	northwest	21984.47061	
6 46 female 33.440 1 no southeast 8240.58960 7 37 female 27.740 3 no northwest 7281.50560 8 37 male 29.830 2 no northeast 6406.41070	4	32	male	28.880	0	no	northwest	3866.85520	
7 37 female 27.740 3 no northwest 7281.50560 8 37 male 29.830 2 no northeast 6406.41070	5	31	female	25.740	0	no	southeast	3756.62160	
8 37 male 29.830 2 no northeast 6406.41070	6	46	female	33.440	1	no	southeast	8240.58960	
	7	37	female	27.740	3	no	northwest	7281.50560	
9 60 female 25.840 0 no northwest 28923.13692	8	37	male	29.830	2	no	northeast	6406.41070	
	9	60	female	25.840	0	no	northwest	28923.13692	

상위 10개 데이터

[labelEncoder] 을 사용하여 성별, 흡연여부,지역과 같은 데이터를 수치로 바꿔준다.

```
from sklearn.preprocessing import LabelEncoder
  #sex
 le = LabelEncoder()
 le.fit(data.sex.drop_duplicates())
 data.sex = le.transform(data.sex)
  # smoker or not
 le.fit(data.smoker.drop_duplicates())
 data.smoker = le.transform(data.smoker)
  #region
 le.fit(data.region.drop_duplicates())
 data.region = le.transform(data.region)
 + Code
             + Markdown
 print(data.region[:10])
 print(le.classes_)
Name: region, dtype: int64
['northeast' 'northwest' 'southeast' 'southwest']
```

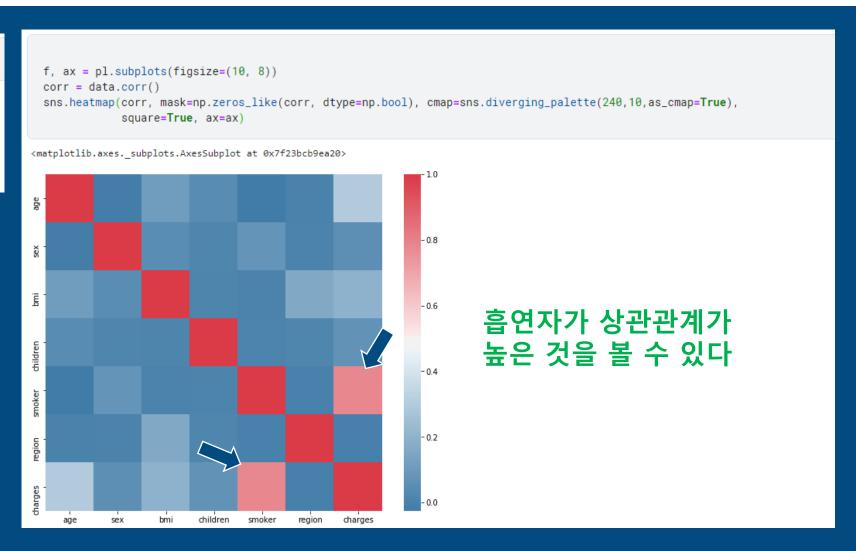
```
print(data.smoker[:10])
      ◆──-흡연자
      ← 비흡연자
5
Name: smoker, dtype: int64
 print(data.sex[:10])
            여성
6
Name: sex, dtype: int64
```

### #3 읽어온 데이터 표시-히트맵

```
data.isnull().sum()

age 0
sex 0
bmi 0
children 0
smoker 0
region 0
charges 0
dtype: int64
```

널값이 있는지 확인



### #3 읽어온 데이터 표시-히스토그램

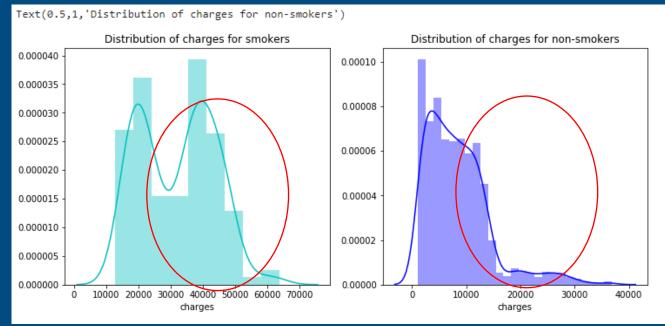
```
from bokeh.io import output_notebook, show
from bokeh.plotting import figure
output_notebook()
import scipy.special
from bokeh.layouts import gridplot
from bokeh.plotting import figure, show, output_file

f= pl.figure(figsize=(12,5))

ax=f.add_subplot(121)
sns.distplot(data[(data.smoker == 1)]["charges"],color='c',ax=ax)
ax.set_title('Distribution of charges for smokers')

ax=f.add_subplot(122)
sns.distplot(data[(data.smoker == 0)]['charges'],color='b',ax=ax)
ax.set_title('Distribution of charges for non-smokers')
```

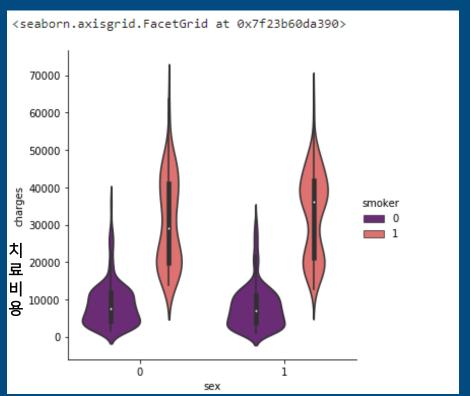
#### 흡연자 vs 비흡연자



#### 흡연자중에 의료비용이 높은 사람이 많다

### #3 읽어온 데이터 표시-catplot, violin

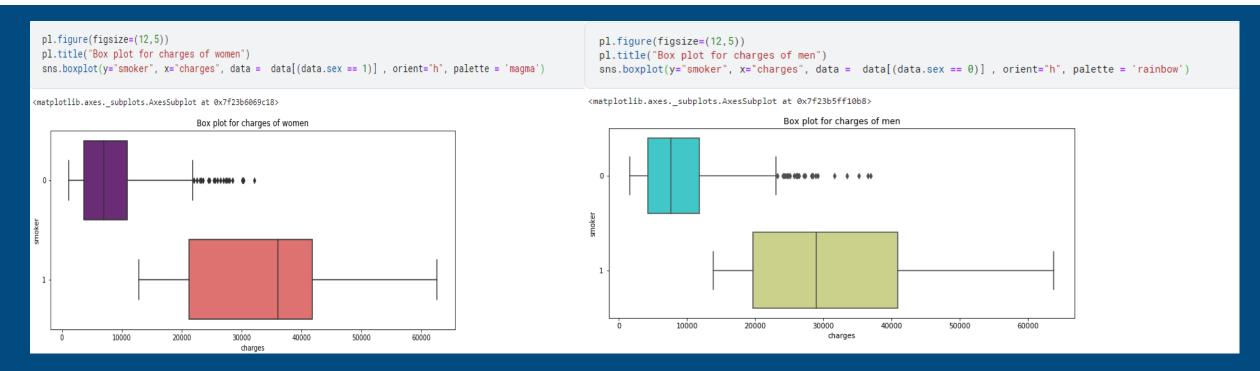




< 성별에 따른 흡연자 > 여성보다 남성흡연자의수가 많다

흡연자가 치료비용이 더 높다

### #3 읽어온 데이터 표시-boxplot

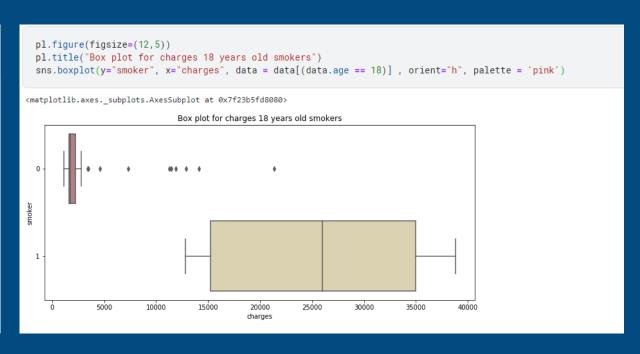


<여성>

<남성>

둘 다 흡연자의 의료 비용의 평균이 더 높다

#### #3 18세의 데이터

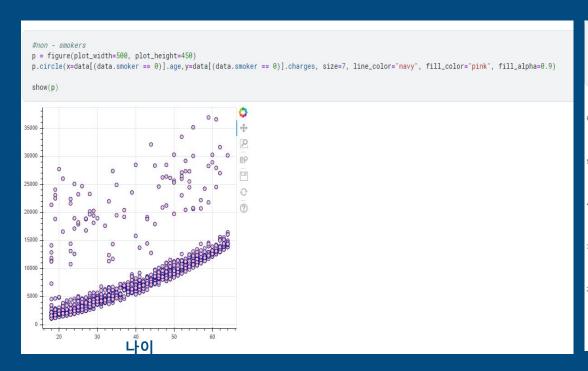


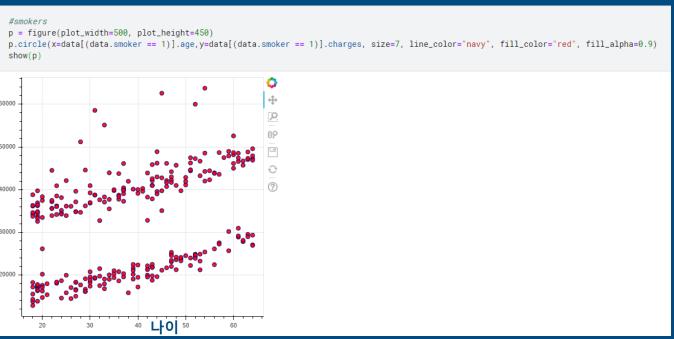
18살 흡연자의 비

⇒ 성인과 비슷

흡연자와 비흡연장의 <u>의료비용 차이가</u> 크다

#### #3 비흡연자 vs 흡연자 – 나이에 따른 의료비용



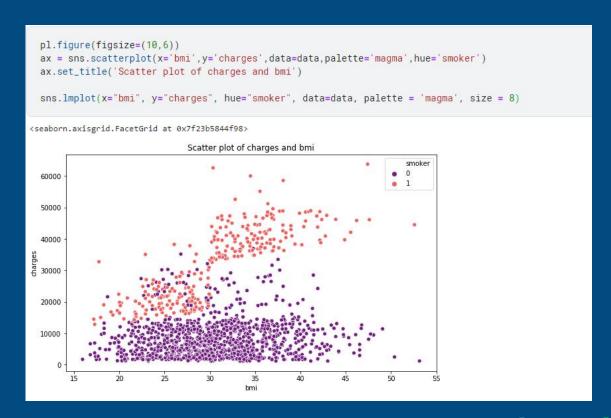


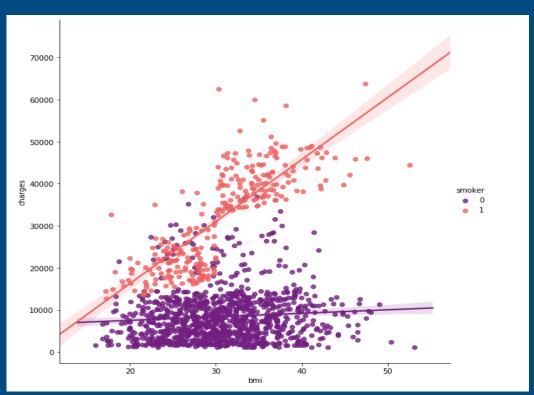
<비흡연자>

<흡연자>

둘 다 나이에 따라 의료비용 증가 흡연자가 의료비용 더 높음

### #3 Bmi(체질량 지수) 에 따른 의료 비용





Bmi보다는 흡연여부가 더 큰 영향을 미침

#### #4 학습 및 테스트(LinearRegression)

```
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import PolynomialFeatures
from sklearn.metrics import r2_score, mean_squared_error
from sklearn.ensemble import RandomForestRegressor
```

```
x = data.drop(['charges'], axis = 1) #문제부분
y = data.charges #답부분

x_train,x_test,y_train,y_test = train_test_split(x,y, random_state = 0)
lr = LinearRegression().fit(x_train,y_train) #학습

y_train_pred = lr.predict(x_train)
y_test_pred = lr.predict(x_test)

print(lr.score(x_test,y_test))
```

0.7962732059725786

#점수

### #4 학습 및 테스트(RandomForestRegressor)

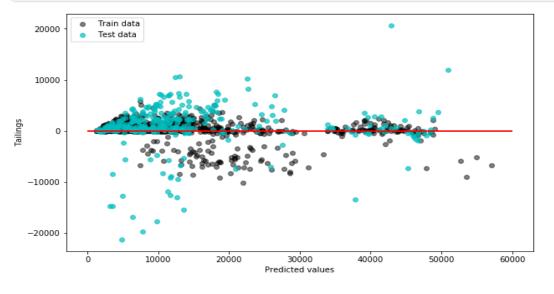
```
RF = RandomForestRegressor(random_state = 0)
RF.fit(x_train,y_train) #학습

score = RF.score(x_train,y_train)
print('Score:', format(score,'.3f'))

pred = RF.predict(x_test)
print('Predicted:', pred)
print('Correct answer:\n', y_test)
```

Score: 0.964 #정수

## 학습용데이터와 테스트데이터가 거의 비슷하다



# 감사합니다.