# output

July 5, 2023

# 1 Data sets for testing the program

# 1.1 1. Predicting Medical Insurance Costs

#### 1.1.1 Summary

From a data set that compiles information on peoples' medical history we implement a linear regression model that attempts to predict the insurance costs of patients.

#### Data Set Description (source)

- 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<sup>2</sup>) 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

#### 1.2 2. Predicting Real Estate Value In The Suburbs Of Boston

#### **1.2.1** Summary

Predict the price of real estate based on different characterizing factors.

#### Data Set Description (source)

- CRIM per capita crime rate by town.
- ZN proportion of residential land zoned for lots over 25,000 sq.ft.
- INDUS proportion of non-retail business acres per town.
- CHAS Charles River dummy variable (= 1 if tract bounds river; 0 otherwise).
- NOX nitrogen oxides concentration (parts per 10 million).
- RM average number of rooms per dwelling.
- AGE proportion of owner-occupied units built prior to 1940.
- DIS weighted mean of distances to five Boston employment centres.
- RAD index of accessibility to radial highways.

- TAX full-value property-tax rate per \$10,000.
- PTRATIO pupil-teacher ratio by town.
- LSTAT lower status of the population (percent).
- MEDV median value of owner-occupied homes in \$1000s.

Control script: ctrl\_insurance.py

#### #### Initial Variables:

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1338 entries, 0 to 1337
Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype
0	age	1338 non-null	int64
1	sex	1338 non-null	object
2	bmi	1338 non-null	float64
3	children	1338 non-null	int64
4	smoker	1338 non-null	object
5	region	1338 non-null	object
6	charges	1338 non-null	float64
dtyp	es: float6	4(2), int64(2),	object(3)

memory usage: 73.3+ KB

None

## Data Viewer 1:

	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	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
9	60	female	25.840	0	no	northwest	28923.13692
10	25	male	26.220	0	no	northeast	2721.32080

Variable Description Before Data Processing:

age bmi children charges count 1338.000000 1338.000000 1338.000000

mean	39.207025	30.663397	1.094918	13270.422265
std	14.049960	6.098187	1.205493	12110.011237
min	18.000000	15.960000	0.000000	1121.873900
25%	27.000000	26.296250	0.000000	4740.287150
50%	39.000000	30.400000	1.000000	9382.033000
75%	51.000000	34.693750	2.000000	16639.912515
max	64.000000	53.130000	5.000000	63770.428010

# Missing Values (NAs) Per Column:

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

# Missing values (NAs) After Replacement/Removal:

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

## #### Variables After Transformation:

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1338 entries, 0 to 1337
Data columns (total 8 columns):

#	Column	Non-Null Count	Dtype
0	age	1338 non-null	int64
1	bmi	1338 non-null	float64
2	children	1338 non-null	int64
3	region	1338 non-null	object
4	charges	1338 non-null	float64
5	sex_d	1338 non-null	int64
6	smoker_d	1338 non-null	int64
7	log_charges	1338 non-null	float64

dtypes: float64(3), int64(4), object(1)

memory usage: 83.8+ KB

None

#### Data Viewer 2:

	age	bmi	children	region	charges	sex_d	smoker_d	log_charges
0	19	27.900	0	southwest	16884.92400	0	0	9.734176
1	18	33.770	1	southeast	1725.55230	1	1	7.453302
2	28	33.000	3	southeast	4449.46200	1	1	8.400538
3	33	22.705	0	northwest	21984.47061	1	1	9.998092
4	32	28.880	0	northwest	3866.85520	1	1	8.260197
5	31	25.740	0	southeast	3756.62160	0	1	8.231275
6	46	33.440	1	southeast	8240.58960	0	1	9.016827
7	37	27.740	3	northwest	7281.50560	0	1	8.893093
8	37	29.830	2	northeast	6406.41070	1	1	8.765054
9	60	25.840	0	northwest	28923.13692	0	1	10.272397
10	25	26.220	0	northeast	2721.32080	1	1	7.908873

## Checking Non-numerical Variables:

- Non-numeric variables in the main data frame:
  - region
- Non-numeric variables dropped.

#### Pearson's Correlations

	log_charges	${\tt smoker\_d}$	sex_d	charges	children	bmi
age	0.528	0.025	-0.021	0.299	0.042	0.109
bmi	0.133	-0.004	0.046	0.198	0.013	_
children	0.161	-0.008	0.017	0.068	_	_
charges	0.893	-0.787	0.057	_	_	-
sex_d	0.006	-0.076	_	_	_	-
smoker_d	-0.666	_	_	_	_	_

# ## Analysis Of Variance Inflation Factor:

Features group: 1

sex\_d 1.000435 age 1.000435

```
Name: VIF, dtype: float64
Features group: 2
sex d 1.002838
     1.012775
age
bmi
      1.014516
Name: VIF, dtype: float64
_____
Features group: 3
age 1.000988
sex_d 1.006202
smoker_d 1.006394
Name: VIF, dtype: float64
_____
Features group: 4
children 1.002242
smoker_d 1.006457
sex_d 1.008878
bmi
        1.014578
         1.015129
age
Name: VIF, dtype: float64
-----
#### Feature Selection Algorithms
## Univariate Selection (Select k Best):
Container index: 1
Parameters: {'target': 'charges', 'k_vars': 4, 'criterion': 'f_regression'}
Variables selected:
      - 'age'
      - 'bmi'
      - 'children'
      - 'smoker_d'
```

```
Container index: 2
Parameters: {'target': 'log_charges', 'k_vars': 4, 'criterion': 'f_regression'}
Variables selected:
       - 'age'
       - 'bmi'
       - 'children'
       - 'smoker_d'
______
## Unique Combinations Of Explanatory Variables Derived From The Feature
Selection Stage:
- Target: 'charges'
   - ['age', 'bmi', 'children', 'smoker_d']
- Target: 'log_charges'
   - ['age', 'bmi', 'children', 'smoker_d']
Total: 2
#### Regression Results
## Summary:
Regression Nr: 1:
   - Selection strategy: univariate
   - Target: 'charges'
   - Explanatory Variables: ['age', 'bmi', 'children', 'smoker_d']
Regression Nr: 2:
   - Selection strategy: univariate
   - Target: 'log_charges'
   - Explanatory Variables: ['age', 'bmi', 'children', 'smoker_d']
Regression Nr: 3:
   - Selection strategy: manually selected
```

- Target: 'charges'
- Explanatory Variables: ['age', 'sex\_d']

## Regression Nr: 4:

- Selection strategy: manually selected
- Target: 'charges'
- Explanatory Variables: ['age', 'sex\_d', 'bmi']

#### Regression Nr: 5:

- Selection strategy: manually selected
- Target: 'charges'
- Explanatory Variables: ['age', 'sex\_d', 'smoker\_d']

#### Regression Nr: 6:

- Selection strategy: manually selected
- Target: 'charges'
- Explanatory Variables: ['age', 'sex\_d', 'bmi', 'children', 'smoker\_d']

#### ### Regression number: 1

Target variable (Y): 'charges'

## Explanatory Variables:

- x1: 'age'
- x2: 'bmi'
- x3: 'children'
- x4: 'smoker\_d'

# Results: Ordinary least squares

Model:		OLS	OLS		squared:	0.741
Depend	ent Variable	э: у		AIC:		21706.1732
Date:		2023-07-	05 16:47	BIC:		21731.0503
No. Ob	servations:	1070		Log-Lik	elihood:	-10848.
Df Mod	el:	4	4		stic:	764.4
Df Res	iduals:	1065	1065		-statistic):	3.77e-311
R-squa	red:	0.742	0.742		Scale:	
	Coef.	Std.Err.	t	P> t	[0.025	0.975]
const	11580.5841	1116.1004	10.3759	0.0000	9390.5787	13770.5895
x1	257.5558	13.5998	18.9383	0.0000	230.8704	284.2412
x2	320.1445	30.8758	10.3688	0.0000	259.5602	380.7288
x3	556.2747	158.0091	3.5205	0.0004	246.2302	866.3193

#### -23597.6939 465.1911 -50.7269 0.0000 -24510.4891 -22684.8987 x4

Omnibus:	221.025	Durbin-Watson:	2.073
<pre>Prob(Omnibus):</pre>	0.000	Jarque-Bera (JB):	455.726
Skew:	1.174	Prob(JB):	0.000
Kurtosis:	5.170	Condition No.:	310

#### Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

#### ## Error measurement:

MSE: 32623008.47 RMSE: 5711.66

### Regression number: 2

Target variable (Y): 'log\_charges'

#### Explanatory Variables:

- x1: 'age' - x2: 'bmi'

- x3: 'children' - x4: 'smoker\_d'

#### Results: Ordinary least squares

\_\_\_\_\_\_ Model: OLS Adj. R-squared: 0.745 1369.0198 Dependent Variable: y AIC: 2023-07-05 16:47 BIC: Date: 1393.8969 Log-Likelihood: No. Observations: 1070 -679.51 783.7 Df Model: F-statistic: Df Residuals: 1065 Prob (F-statistic): 1.98e-315 R-squared: 0.746

	Coef.	Std.Err.	t	P> t	[0.025	0.975]
const x1 x2	8.5427 0.0340 0.0107	0.0833 0.0010 0.0023	102.5989 33.5202 4.6477	0.0000 0.0000 0.0000	8.3794 0.0320 0.0062	8.7061 0.0360 0.0152
x3	0.1016	0.0118	8.6210	0.0000	0.0785	0.1248
x4	-1.5172 	0.0347	-43.7170 	0.0000	-1.5853 	-1.4491

Durbin-Watson: 2.020 Omnibus: 347.799 Prob(Omnibus): 0.000 Jarque-Bera (JB): 1143.341 

 Skew:
 1.589
 Prob(JB):
 0.000

 Kurtosis:
 6.943
 Condition No.:
 310

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#### Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

#### ## Error measurement:

MSE: 0.18 RMSE: 0.42

### Regression number: 3

Target variable (Y): 'charges'

#### Explanatory Variables:

- x1: 'age' - x2: 'sex d'

#### Results: Ordinary least squares

\_\_\_\_\_\_ Model: OLS. Adj. R-squared: 0.088 AIC: Dependent Variable: y 23049.7374 23064.6636 Date: 2023-07-05 16:47 BIC: No. Observations: 1070 Log-Likelihood: -11522.

Df Model: 2 F-statistic: 52.66

Df Residuals: 1067 Prob (F-statistic): 1.55e-22

R-squared: 0.090 Scale: 1.3237e+0 1.3237e+08 \_\_\_\_\_ Std.Err. t P>|t| [0.025 Coef. 0.975] \_\_\_\_\_\_ const 2605.5835 1128.9424 2.3080 0.0212 390.3843 4820.7826 255.4248 25.2778 10.1047 0.0000 205.8251 305.0246 x1 1462.8520 703.7863 2.0785 0.0379 81.8898 2843.8142 \_\_\_\_\_ 292.892 Durbin-Watson: Omnibus: 1.933 Jarque-Bera (JB): 581.679 Prob(Omnibus): 0.000 Skew: 1.642 Prob(JB): 0.000 4.506 Kurtosis: Condition No.: 139

Notes:

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#### ## Error measurement:

<sup>[1]</sup> Standard Errors assume that the covariance matrix of the errors is correctly specified.

MSE: 128284398.24 RMSE: 11326.27

### Regression number: 4

Target variable (Y): 'charges'

#### Explanatory Variables:

- x1: 'age' - x2: 'sex\_d' - x3: 'bmi'

#### Results: Ordinary least squares

\_\_\_\_\_\_ OLS Adj. R-squared: Model: 0.113 Dependent Variable: y AIC: 23021.0051 Date: 2023-07-05 16:47 BIC: 23040.9068 Log-Likelihood: -11507. F-statistic: 46.45 No. Observations: 1070 3 F-statistic: 46.45 Prob (F-statistic): 3.26e-28 Df Model: 1066 Df Residuals: 0.116 Scale: 1.2875e+08 R-squared: \_\_\_\_\_ Std.Err. t P>|t| [0.025 0.975Coef. \_\_\_\_\_ const -6373.0901 1958.3127 -3.2544 0.0012 -10215.6752 -2530.5049 238.2345 25.1191 9.4842 0.0000 188.9459 x1 x2 1199.9390 695.6781 1.7248 0.0848 -165.1149 2564.9929 318.9589 57.2301 5.5733 0.0000 206.6624 431.2554 233.907 1.933 Omnibus: Durbin-Watson: Jarque-Bera (JB): Prob(Omnibus): 403.213 0.000 Skew: Prob(JB): 1.434 0.000 3.907 Condition No.: Kurtosis: 292 \_\_\_\_\_\_

#### Notes

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

# ## Error measurement:

MSE: 126155971.63 RMSE: 11231.92

### Regression number: 5

Target variable (Y): 'charges'

# Explanatory Variables:

- x1: 'age' - x2: 'sex\_d' - x3: 'smoker\_d'

## Results: Ordinary least squares

\_\_\_\_\_

Model: Dependent Variable: Date: No. Observations: Df Model: Df Residuals: R-squared:		2023-07-05 16:47 1070 3 1066		Adj. R-squared: AIC: BIC: Log-Likelihood: F-statistic: Prob (F-statistic): Scale:		21819.0882 21838.9899 -10906. 880.1 7.72e-288
	Coef.	Std.Err.	t	P> t	[0.025	0.975]
x1 x2	277.1770 42.4621	14.2235 396.9149	19.4872 0.1070	0.0000 0.9148	19795.1941 249.2677 -736.3612 -24595.7476	305.0862 821.2854
Omnibus: Prob(Omnibus): Skew: Kurtosis:		0.000 . 1.053 I		Durbin-Watson: Jarque-Bera (JB): Prob(JB): Condition No.:		2.072 413.935 0.000 173

# Notes:

 $\[1\]$  Standard Errors assume that the covariance matrix of the errors is correctly specified.

#### ## Error measurement:

MSE: 36000763.69 RMSE: 6000.06

### Regression number: 6

Target variable (Y): 'charges'

# Explanatory Variables:

- x1: 'age'

- x2: 'sex\_d' - x3: 'bmi'

- x4: 'children' - x5: 'smoker\_d'

# Results: Ordinary least squares

\_\_\_\_\_

Model:		OLS		Adj. R-squared:		0.741
Dependent Variable:		: y		AIC:		21707.6711
Date:		2023-07-05 16:47		BIC:		21737.5236
No. Observations:		1070		Log-Likelihood:		-10848.
Df Model:		5		F-statistic:		611.4
Df Residuals:		1064		<pre>Prob (F-statistic):</pre>		8.65e-310
R-squared:		0.742		Scale:		3.7659e+07
	Coef.		t 		[0.025	0.975]
const	11699.2207				9484.0686	13914.3727
x1	257.2126	13.6116	18.8965	0.0000	230.5039	283.9212
x2	-266.7664	377.4882	-0.7067	0.4799	-1007.4724	473.9395
x3	321.6202	30.9536	10.3904	0.0000	260.8832	382.3572
x4	559.8364	158.1266	3.5404	0.0004	249.5610	870.1119
x5	-23622.1141	466.5819	-50.6280	0.0000	-24537.6392	-22706.5890
Omnibus:		220.123 I		Durbin-Watson:		2.075
				Jarque-Bera (JB):		
Skew:				Prob(JB):		0.000
Kurtosis:				Condition No.:		314

#### Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

# ## Error measurement:

MSE: 32669542.5 RMSE: 5715.73

# \*\* [No more experiments] \*\*

## [End Of Report]

[NbConvertApp] Making directory

../project\_13/exported\_pdf/notebook\_version\_new\_version.pdf
[NbConvertApp] Converting notebook ../project\_13/output.ipynb to pdf

[NbConvertApp] Writing 40065 bytes to notebook.tex

```
[NbConvertApp] Building PDF
[NbConvertApp] Running xelatex 3 times: ['xelatex', 'notebook.tex', '-quiet']
[NbConvertApp] Running bibtex 1 time: ['bibtex', 'notebook']
[NbConvertApp] WARNING | bibtex had problems, most likely because there were no citations
[NbConvertApp] PDF successfully created
[NbConvertApp] Writing 45093 bytes to
../project_13/exported_pdf/notebook_version_new_version.pdf/output.pdf
```