

16_feature_selection_vjezba

August 1, 2025

```
[1]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import statsmodels.api as sm
from statsmodels.formula.api import ols
```

```
[2]: lifeexp = pd.read_csv("../..../datasets/life_exp.csv")
```

1 Feature selection vježba

Korištenjem feature selection metoda odredite koje varijable bi vam bile korisne pri prognozi životne vijeka Life_Expectancy

Dataset nema missing values kao ni outliers tako da se s time ne moramo zamarati

```
[3]: lifeexp
```

```
[3]:   Unnamed: 0  Life_Expectancy  Average_Temperature  Daily_Screen_Time  \
0           0      72.483571           17.923146           4.357787
1           1      69.308678           22.896773           4.560785
2           2      73.238443           23.286427           5.083051
3           3      77.615149           20.988614           5.053802
4           4      68.829233           24.193571           2.622331
..          ...            ...                ...                ...
95          95      62.682425           26.926587           3.307090
96          96      71.480601           20.580713           4.899600
97          97      71.305276           25.768626           4.307300
98          98      70.025567           25.291044           4.812862
99          99      68.827064           19.285149           4.629629
```

```
   Health_Index  Income  Country  Diet_Type  \
0    71.654576  74294.715447  Country B  Vegetarian
1    68.748497  73109.928448  Country B  Non-Vegetarian
2    73.985736  78248.930090  Country C  Vegetarian
3    78.225520  82709.110470  Country C  Vegan
4    68.808332  72929.102183  Country A  Vegetarian
..          ...            ...                ...
95    62.213250  68760.245347  Country C  Non-Vegetarian
```

96	69.767467	74406.109077	Country A	Non-Vegetarian
97	72.659149	75924.599005	Country B	Non-Vegetarian
98	69.911027	73274.330776	Country C	Non-Vegetarian
99	70.064881	71061.464871	Country C	Vegan

Government_Health_Policy	
0	1.0
1	1.0
2	1.0
3	1.0
4	1.0
..	...
95	1.0
96	1.0
97	1.0
98	1.0
99	1.0

[100 rows x 9 columns]

```
[4]: lifeexp.describe()
```

	Unnamed: 0	Life_Expectancy	Average_Temperature	Daily_Screen_Time \
count	100.000000	100.000000	100.000000	100.000000
mean	49.500000	69.480767	25.111523	4.064896
std	29.011492	4.540842	4.768345	1.084283
min	0.000000	56.901274	15.406144	0.758733
25%	24.750000	66.995472	20.971697	3.344556
50%	49.500000	69.365219	25.420536	4.097696
75%	74.250000	72.029760	27.690852	4.704437
max	99.000000	79.261391	38.600846	7.852731

	Health_Index	Income	Government_Health_Policy
count	100.000000	100.000000	100.0
mean	69.587608	74368.758472	1.0
std	4.475948	4744.899489	0.0
min	59.054457	62441.446464	1.0
25%	66.820378	70968.995974	1.0
50%	69.626849	74399.904642	1.0
75%	72.651242	78126.987900	1.0
max	79.173729	86449.739694	1.0

```
[5]: lifeexp.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100 entries, 0 to 99
Data columns (total 9 columns):
#   Column                                Non-Null Count  Dtype
#   ...
```

```

---      -----      -----      -----
0  Unnamed: 0          100 non-null    int64
1  Life_Expectancy     100 non-null    float64
2  Average_Temperature 100 non-null    float64
3  Daily_Screen_Time   100 non-null    float64
4  Health_Index        100 non-null    float64
5  Income              100 non-null    float64
6  Country             100 non-null    object
7  Diet_Type           100 non-null    object
8  Government_Health_Policy 100 non-null    float64
dtypes: float64(6), int64(1), object(2)
memory usage: 7.2+ KB

```

```
[6]: lifeexp.corr(numeric_only = True)
```

```

[6]:
      Unnamed: 0  Life_Expectancy  Average_Temperature \
Unnamed: 0      1.000000      0.044508      0.006198
Life_Expectancy  0.044508      1.000000      -0.136422
Average_Temperature  0.006198     -0.136422      1.000000
Daily_Screen_Time -0.089520      0.190840     -0.036632
Health_Index      0.043373      0.980875     -0.141879
Income            0.029239      0.895932     -0.045018
Government_Health_Policy  NaN          NaN          NaN

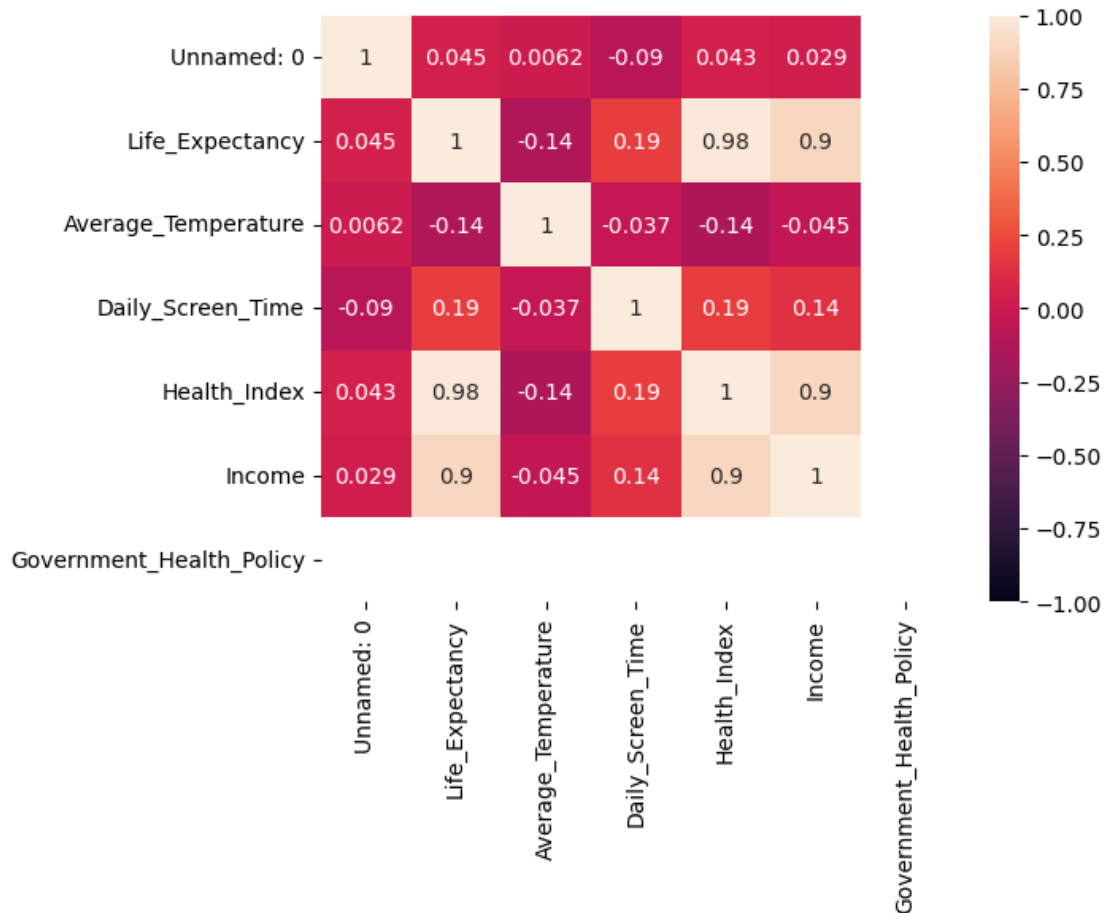
      Daily_Screen_Time  Health_Index  Income \
Unnamed: 0      -0.089520      0.043373  0.029239
Life_Expectancy      0.190840      0.980875  0.895932
Average_Temperature   -0.036632     -0.141879 -0.045018
Daily_Screen_Time      1.000000      0.193556  0.135069
Health_Index           0.193556      1.000000  0.895719
Income                 0.135069      0.895719  1.000000
Government_Health_Policy  NaN          NaN          NaN

      Government_Health_Policy
Unnamed: 0      NaN
Life_Expectancy  NaN
Average_Temperature  NaN
Daily_Screen_Time  NaN
Health_Index      NaN
Income            NaN
Government_Health_Policy  NaN

```

```
[7]: sns.heatmap(lifeexp.corr(numeric_only = True), vmin=-1, vmax=1, annot=True)
```

```
[7]: <Axes: >
```



```
[8]: model = ols('Life_Expectancy ~ C(Country) + C(Diet_Type) + C(Country):
↳C(Diet_Type)', data=lifexp).fit()
```

```
[9]: anova_table = sm.stats.anova_lm(model, typ=2)
```

```
[10]: anova_table
```

```
[10]:
```

	sum_sq	df	F	PR(>F)
C(Country)	28.499972	2.0	0.706552	0.496028
C(Diet_Type)	71.087330	2.0	1.762349	0.177448
C(Country):C(Diet_Type)	115.615265	4.0	1.433128	0.229411
Residual	1835.319446	91.0	NaN	NaN

1.1 Rješenje:

Mičemo sve varijable osim Income

Korelacijska metoda te Anova analiza su nam pokazale kako nema utjecaja tih varijabli na Life_expectancy

Naglasak: ovo je rezultat analize na podacima koje imamo

provjera - napraviti ćemo boxplotove gdje ćemo usporediti zemlju i tip prehrane s očekivanim životnim vijekom...

```
[11]: # Postavljanje stila za plotove
sns.set_style("whitegrid")

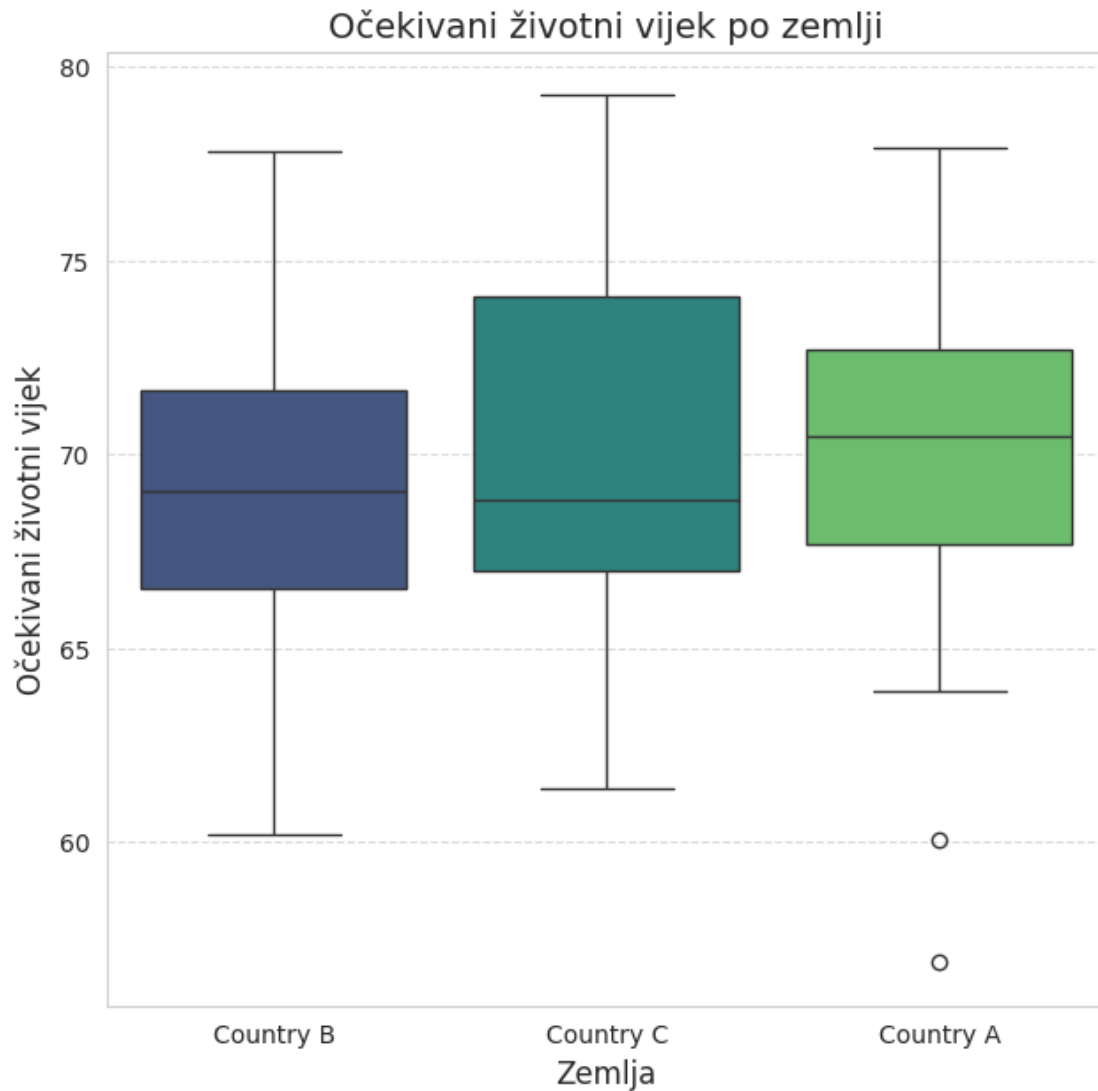
plt.figure(figsize=(16, 7)) # Povećavamo veličinu figure za bolji prikaz

# Box plot za Life_Expectancy po Country
plt.subplot(1, 2, 1) # 1 red, 2 stupca, prvi plot
sns.boxplot(x='Country', y='Life_Expectancy', data=lifeexp, palette='viridis')
plt.title('Očekivani životni vijek po zemlji', fontsize=14)
plt.xlabel('Zemlja', fontsize=12)
plt.ylabel('Očekivani životni vijek', fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7) # Dodajemo mrežu na y-osi
```

/tmp/ipykernel_1509/2779546637.py:8: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.boxplot(x='Country', y='Life_Expectancy', data=lifeexp, palette='viridis')
```



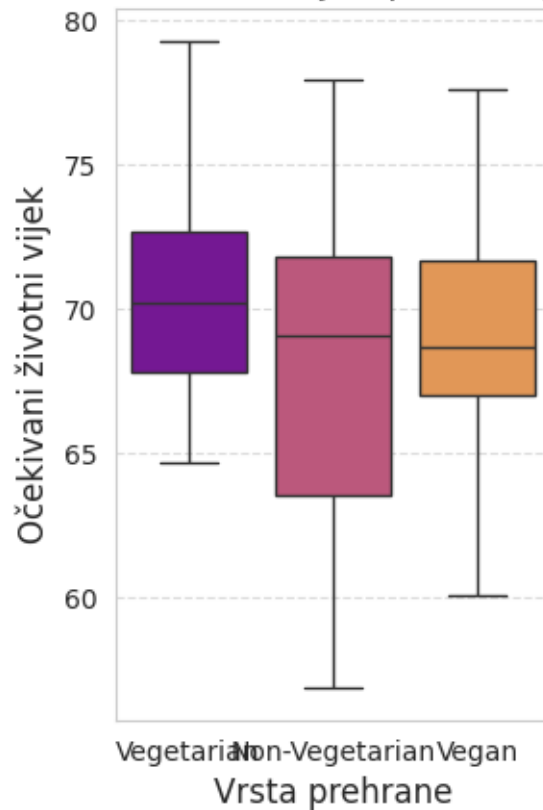
```
[12]: # Box plot za Life_Expectancy po Diet_Type
plt.subplot(1, 2, 2) # 1 red, 2 stupca, drugi plot
sns.boxplot(x='Diet_Type', y='Life_Expectancy', data=lifeexp, palette='plasma')
plt.title('Očekivani životni vijek po vrsti prehrane', fontsize=14)
plt.xlabel('Vrsta prehrane', fontsize=12)
plt.ylabel('Očekivani životni vijek', fontsize=12)
plt.grid(axis='y', linestyle='--', alpha=0.7) # Dodajemo mrežu na y-osi
```

/tmp/ipykernel_1509/334417189.py:3: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.boxplot(x='Diet_Type', y='Life_Expectancy', data=lifeexp,
palette='plasma')
```

Očekivani životni vijek po vrsti prehrane



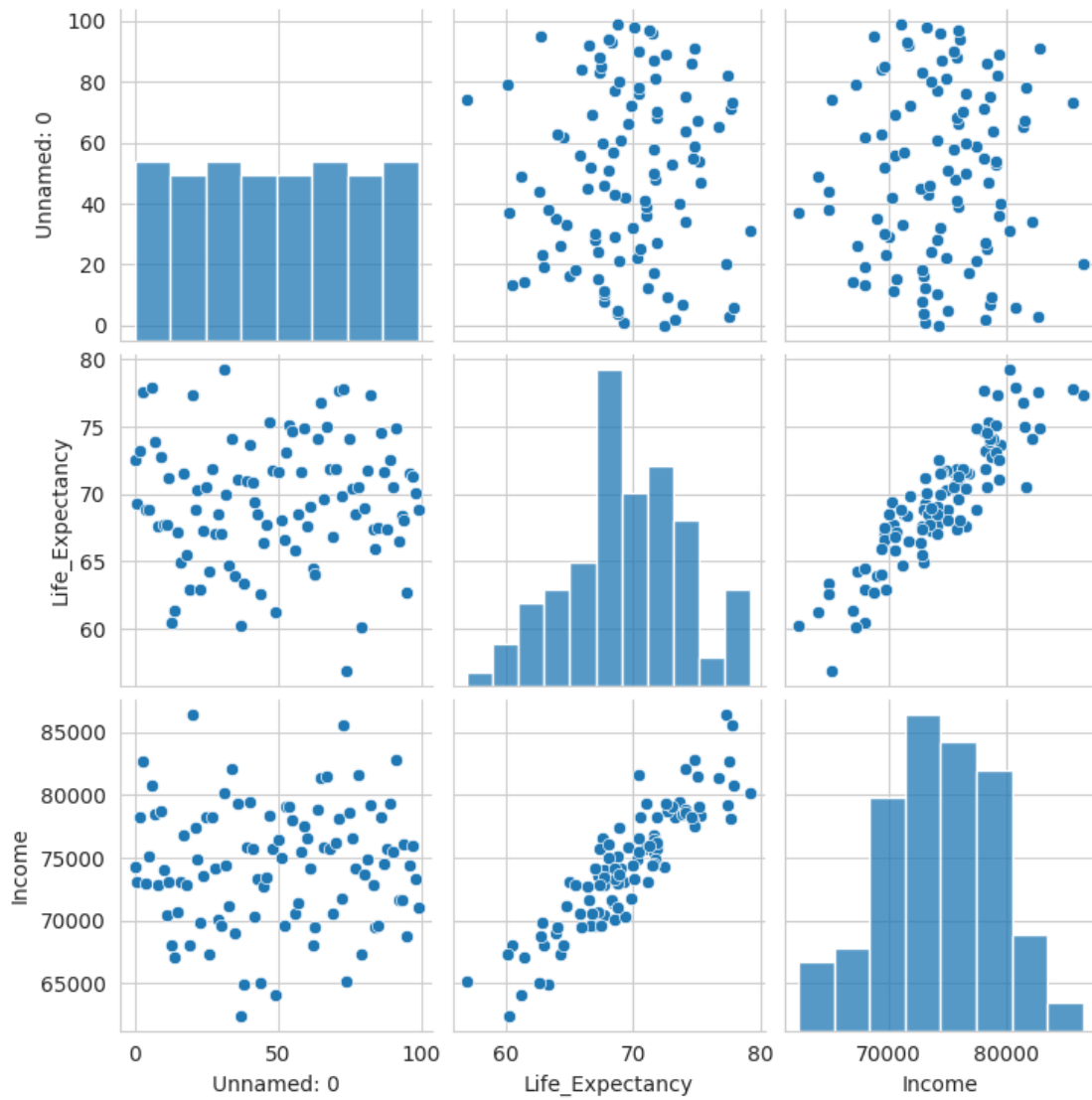
```
[13]: lifeexp.drop(columns=['Government_Health_Policy', 'Daily_Screen_Time',
↪ 'Country', 'Diet_Type', 'Average_Temperature', 'Health_Index'],
↪ inplace=True)
```

```
[14]: lifeexp.head()
```

```
[14]: Unnamed: 0  Life_Expectancy      Income
0           0      72.483571  74294.715447
1           1      69.308678  73109.928448
2           2      73.238443  78248.930090
3           3      77.615149  82709.110470
4           4      68.829233  72929.102183
```

```
[15]: sns.pairplot (lifeexp)
```

```
[15]: <seaborn.axisgrid.PairGrid at 0x7fe812fc28d0>
```



[]: