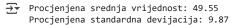
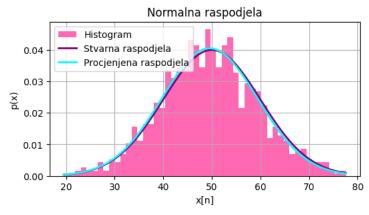
```
# reset svih varijabli u okruženju
from IPython import get_ipython
get_ipython().magic('reset -sf')
```

```
# ZADATAK 1
import numpy as np
from scipy.optimize import minimize
import matplotlib.pyplot as plt
np.random.seed(0)
n=1000
X=np.random.normal(50, 10, n)
def log_vjerovatnoca(params):
    mi, sigma=params[0], params[1]
    return 0.5*n*np.log(2*np.pi)+n*np.log(sigma)+(1/(2*sigma**2))*np.sum((X-mi)**2)
pocetna\_pretpostavka = [np.mean(X), np.std(X)]
rez=minimize(log_vjerovatnoca, pocetna_pretpostavka)
theta=rez.x
print("Procjenjena srednja vrijednost: %.2f\nProcjenjena standardna devijacija: %.2f" % (theta[0], theta[1]))
xmin, xmax=np.min(X), np.max(X)
x=np.linspace(xmin, xmax, 100)
p1=(1/(10*np.sqrt(2*np.pi)))*np.exp(-0.5*((x-50)/10)**2)
p2=(1/(theta[1]*np.sqrt(2*np.pi)))*np.exp(-0.5*((x-theta[0])/theta[1])**2)
plt.figure(figsize=(6, 3))
plt.hist(X, bins=50, density=True, color='hotpink', label='Histogram')
plt.plot(x, p1, linewidth=2, color='purple', label='Stvarna raspodjela')
plt.plot(x, p2, linewidth=2, color='cyan', label='Procjenjena raspodjela')
plt.grid()
plt.title('Normalna raspodjela')
plt.xlabel('x[n]')
plt.ylabel('p(x)')
plt.legend()
from google.colab import files
plt.savefig("z1.png")
files.download("z1.png")
```





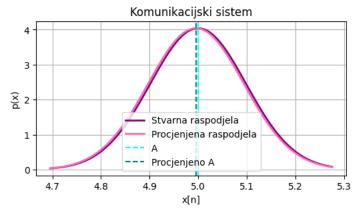
```
# ZADATAK 2 - https://www.gaussianwaves.com/2012/10/likelihood-function-and-maximum-likelihood-estimation-mle/
import numpy as np
import matplotlib.pyplot as plt
from scipy.optimize import minimize

np.random.seed(0)
A=5
n=1000
x=A+0.1*np.random.randn(n)
sigma=np.std(x)

def log_vjerovatnoca(mi):
    return 0.5*n*np.log(2*np.pi)+n*np.log(sigma)+(1/(2*sigma**2))*np.sum((x-mi)**2)
```

```
pocetna_pretpostavka=[np.mean(x)]
rez=minimize(log_vjerovatnoca, pocetna_pretpostavka)
theta=rez.x
print("Procjena DC komponente: %.2f" % (theta[0]))
xmin, xmax = np.min(x), np.max(x)
x=np.linspace(xmin, xmax, 100)
p1=(1/(sigma*np.sqrt(2*np.pi)))*np.exp(-0.5*((x-A)/sigma)**2)
p2=(1/(sigma*np.sqrt(2*np.pi)))*np.exp(-0.5*((x-theta[0])/sigma)**2)
plt.figure(figsize=(6, 3))
plt.plot(x, p1, linewidth=2, color='purple', label='Stvarna raspodjela')
plt.plot(x, p2, linewidth=2, color='hotpink', label='Procjenjena raspodjela')
plt.axvline(A, color='cyan', linestyle='--', label="A")
plt.axvline(theta[0], color='teal', linestyle='--', label="Procjenjeno A")
plt.grid()
plt.title('Komunikacijski sistem')
plt.xlabel('x[n]')
plt.ylabel('p(x)')
plt.legend()
from google.colab import files
plt.savefig("z23.png")
files.download("z23.png")
```

→ Procjena DC komponente: 5.00

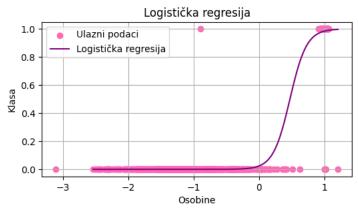


```
# ZADATAK 3 - https://learningdaily.dev/understanding-maximum-likelihood-estimation-in-machine-learning-22b915c3e05a
from sklearn.datasets import make_classification
from sklearn.model_selection import train_test_split
import numpy as np
from scipy.optimize import minimize
import matplotlib.pyplot as plt
n=100
X, Y=make_classification(n_samples=n, n_features=1, n_informative=1, n_redundant=0, n_clusters_per_class=1)
X_trening, X_test, Y_trening, Y_test=train_test_split(X, Y, test_size=0.2, random_state=0)
def log_vjerovatnoca(theta, X, y):
    z=np.dot(X, theta)
    return -np.sum(y*z-np.log(1+np.exp(z)))
theta_pocetno=np.zeros(X_trening.shape[1]+1)
X_modifikovano=np.c_[np.ones(X_trening.shape[0]), X_trening]
rez=minimize(log_vjerovatnoca, theta_pocetno, args=(X_modifikovano, Y_trening))
x_modifikovano=np.c_[np.ones(X_test.shape[0]), X_test]
p=1/(1+np.exp(-np.dot(x_modifikovano, theta)))
p=(p>=0.5).astype(int)
tacnost=np.mean(p==Y test)
print("Tačnost na testnom skupu podataka: ", tacnost)
x=np.linspace(min(X_trening), max(X_trening), n)
x_{modifikovano=np.c_{np.ones}(x.shape[0]), x]
p=1/(1+np.exp(-np.dot(x_modifikovano, theta)))
plt.figure(figsize=(6, 3))
plt.scatter(X, Y, color='hotpink', label='Ulazni podaci')
plt.plot(x, p, color='purple', label='Logistička regresija')
```

```
plt.grid()
plt.xlabel('Osobine')
plt.ylabel('Klasa')
plt.title('Logistička regresija')
plt.legend()

from google.colab import files
plt.savefig("z3.png")
files.download("z3.png")
```

Tačnost na testnom skupu podataka: 0.99



```
# ZADATAK 4 - https://python.quantecon.org/mle.html
import numpy as np
from scipy.special import factorial
from statsmodels.api import Poisson
def log_vjerovatnoca(X, y, beta):
    mi=np.exp(np.dot(X, beta))
    return np.sum(y*np.log(mi)-mi-np.log(factorial(y)))
def newton_raphson(X, y, beta_pocetno):
    print(f'\{"Iteracija":<13\}\{"Log-vjerovatnoća":<18\}\{"\theta"\}')
    print("-" * 56)
    greska=100
    beta=beta_pocetno.reshape(-1, 1)
    y=y.reshape(X.shape[0], 1)
    i=0
    while np.any(greska>1e-3) and i<1000:
        mi=np.exp(np.dot(X, beta))
        H=-np.dot(X.T, mi*X)
        G=np.dot(X.T, y-mi)
        beta_1=beta-np.dot(np.linalg.inv(H), G)
        greska=np.abs(beta_1-beta)
        beta=beta_1
        \label{log_vjerovatnoca} print(f'\{i:<13\}\{log\_vjerovatnoca(X, y, beta):<18.8\}\{beta.flatten()\}')
    print('Broj iteracija: %d\nβ_hat:' % (i))
    print(beta.flatten())
    return beta.flatten()
X=np.array([[1, 2, 5], [1, 1, 3], [1, 4, 2], [1, 5, 2], [1, 3, 1]])
Y=np.array([1, 0, 1, 1, 0])
\texttt{beta\_pocetno=np.array}([0.1,\ 0.1,\ 0.1])
newton_raphson(X, Y, beta_pocetno=beta_pocetno)
print("\n")
print(Poisson(Y, X).fit().summary())
```

```
Titeracija Log-vjerovatnoća θ

0 -4.3447622 [-1.48899977 0.26480369 0.24381722]
1 -3.5742413 [-3.3839381 0.52781212 0.47408367]
2 -3.3999526 [-5.0644346 0.78159328 0.70175342]
3 -3.3788646 [-5.91511889 0.90916891 0.82034123]
```

```
4
            -3.3783559
                             [-6.07389901 0.93272496 0.84265078]
5
            -3.3783555
                              [-6.07848205 0.93340226 0.84329625]
                              [-6.07848573 0.9334028 0.84329677]
            -3.3783555
Broj iteracija: 7
β_hat:
[-6.07848573 0.9334028 0.84329677]
Optimization terminated successfully.
         Current function value: 0.675671
        Iterations 7
                        Poisson Regression Results
Dep. Variable:
                                  y No. Observations:
                   Poisson Df Residuals:
MLE Df Model:
Model:
Method:
Date:
                  Mon, 20 Jan 2025 Pseudo R-squ.:
                                                                    0.2546
                   18:46:54 Log-Likelihood:
True LL-Null:
                                                                     -3.3784
Time:
converged:
converged: True LL-Null:
Covariance Type: nonrobust LLR p-value:
                                                                     -4.5325
______
             coef std err z P>|z| [0.025 0.975]
        -6.0785

    -6.0785
    5.279
    -1.151
    0.250
    -16.425

    0.9334
    0.829
    1.126
    0.260
    -0.691

    0.8433
    0.798
    1.057
    0.291
    -0.720

const
                                                                       2.558
x1
                                                                       2.407
```

```
# ZADATAK 5 - https://python.quantecon.org/mle.html
import pandas as pd
from statsmodels.api import Poisson
from statsmodels.iolib.summary2 import summary_col
import matplotlib.pyplot as plt
\label{lem:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def-pd:def
df=df[df['year']==2008]
df['const']=1
r1=['const', 'lngdppc', 'lnpop', 'gattwto08']
r2=['const', 'lngdppc', 'lnpop', 'gattwto08', 'lnmcap08', 'rintr', 'topint08']
r3=['const', 'lngdppc', 'lnpop', 'gattwto08', 'lnmcap08', 'rintr', 'topint08', 'nrrents', 'roflaw']
r=[r1, r2, r3]
r_imena=['Model 1', 'Model 2', 'Model 3']
rez=[]
for i in r:
        result=Poisson(df[['numbil0']], df[i], missing='drop').fit()
        rez.append(result)
tabela=summary_col(results=rez, float_format='%0.3f', stars=True, model_names=r_imena)
print(tabela)
data=['const', 'lngdppc', 'lnpop', 'gattwto08', 'lnmcap08', 'rintr', 'topint08', 'nrrents', 'roflaw', 'numbil0', 'country']
rez_poredba=df[data].dropna()
rez_poredba['prediction']=rez[2].predict()
rez_poredba['difference']=rez_poredba['numbil0']-rez_poredba['prediction']
rez_poredba.sort_values('difference', ascending=False, inplace=True)
rez_poredba[:27].plot('country', 'difference', kind='bar', figsize=(6,3), legend=False, color='purple')
plt.title('Broj milijardera preko predviđenog broja')
plt.xlabel('Država')
print('\nUkupan broj država:', rez_poredba['country'].nunique())
print('Razlika > %d: %d' % (prag, (rez_poredba['difference']>prag).sum()))
print('Razlika je ±%d: %d' % (prag, ((rez_poredba['difference']>=-prag) & (rez_poredba['difference']<=prag)).sum()))</pre>
print('Maksimalna greška: %d' % rez_poredba['difference'].max())
print('Minimalna greška: %d' % rez_poredba['difference'].min())
from google.colab import files
plt.savefig("z5.png")
files.download("z5.png")
```

Optimization terminated successfully.

Current function value: 2.226090

Iterations 9

Optimization terminated successfully.

Current function value: 1.982676

Iterations 9

Optimization terminated successfully.

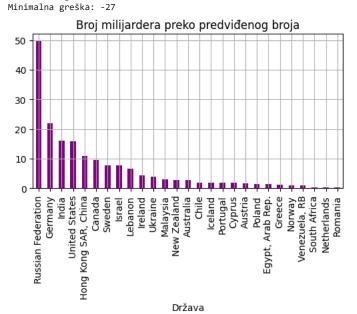
Current function value: 1.954384

Iterations 9

=======			
	Model 1	Model 2	Model 3
const	-29.050***	-19.444***	-20.858***
	(0.638)	(1.668)	(1.742)
1ngdppc	1.084***	0.717***	0.737***
	(0.035)	(0.083)	(0.084)
lnpop	1.171***	0.806***	0.929***
	(0.024)	(0.071)	(0.085)
gattwto08	0.006***	0.007***	0.004*
	(0.002)	(0.002)	(0.002)
lnmcap08		0.399***	0.286***
		(0.059)	(0.071)
rintr		-0.010**	-0.009*
		(0.004)	(0.004)
topint08		-0.051***	-0.058***
		(0.003)	(0.005)
nrrents			-0.005
			(0.005)
roflaw			0.203**
			(0.100)

Standard errors in parentheses.

Ukupan broj država: 131 Razlika > 1: 22 Razlika je ±1: 83 Razlika < 1: 109 Maksimalna greška: 49



^{*} p<.1, ** p<.05, ***p<.01