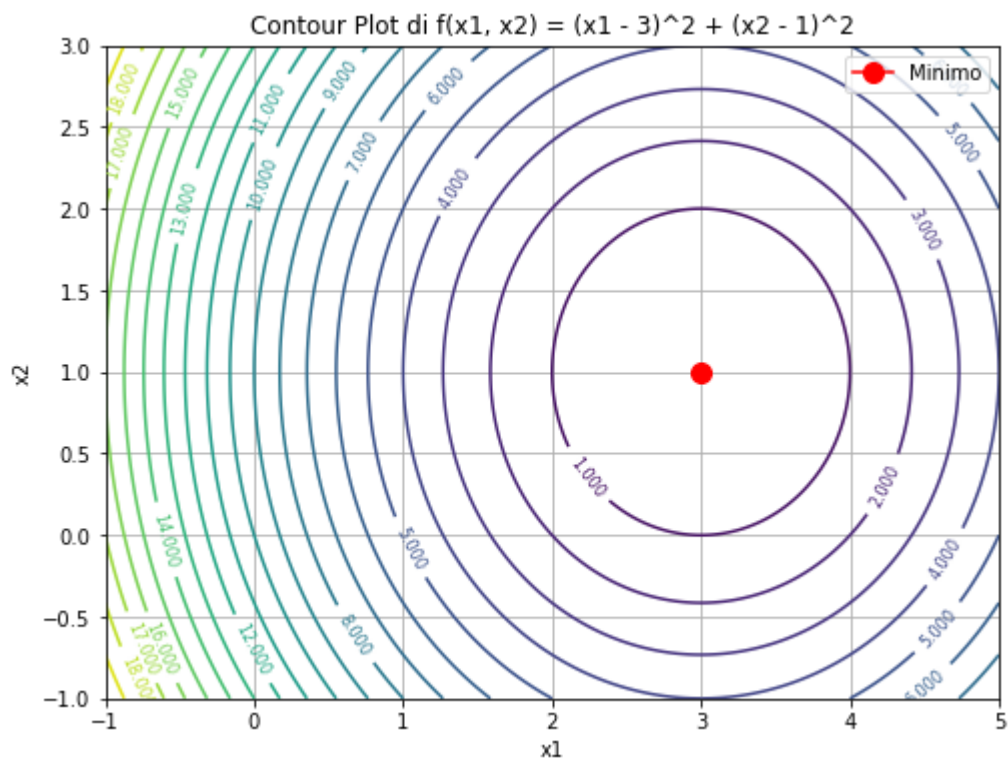
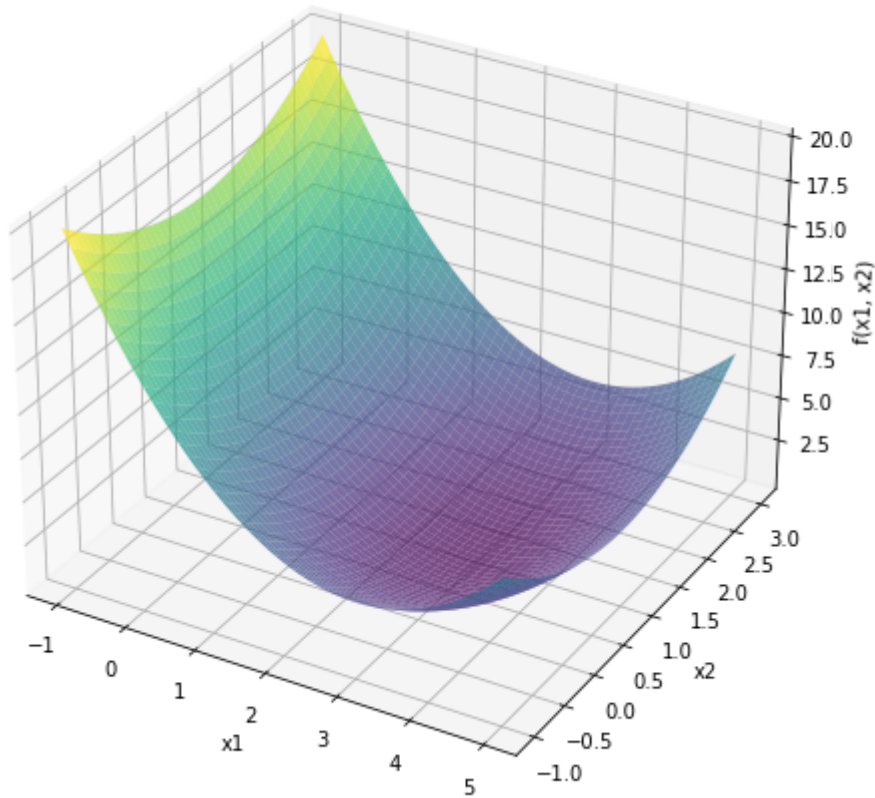
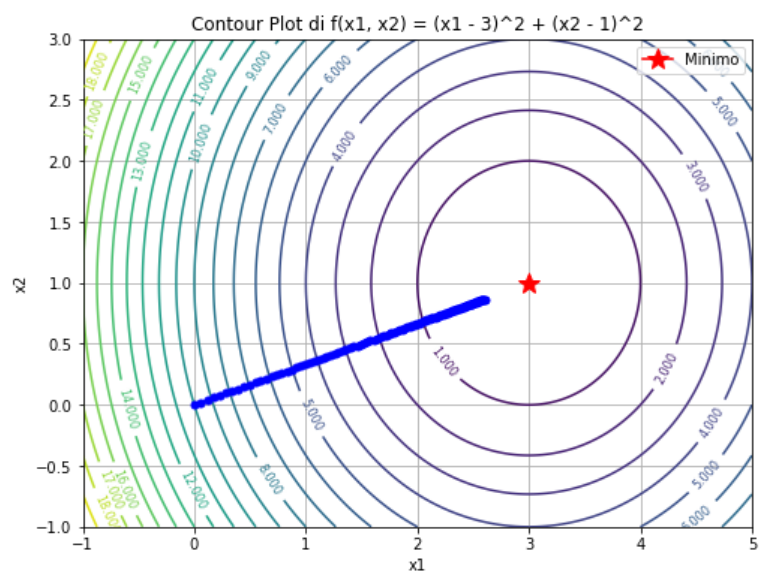
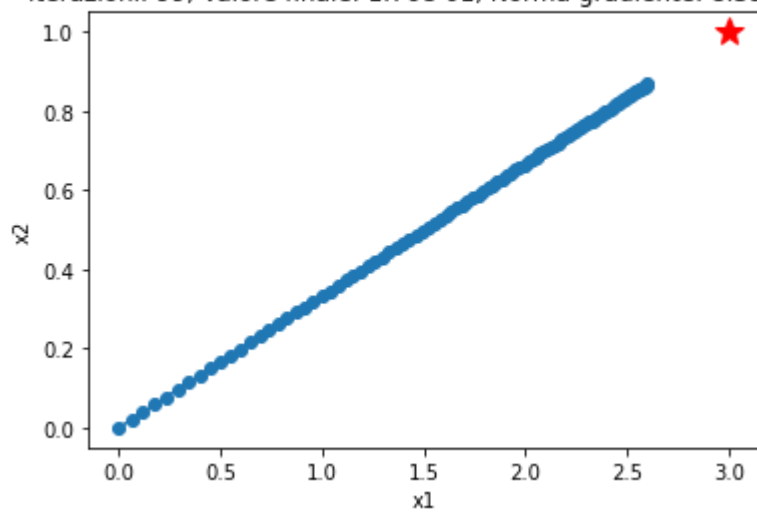


HomeWork 3_1

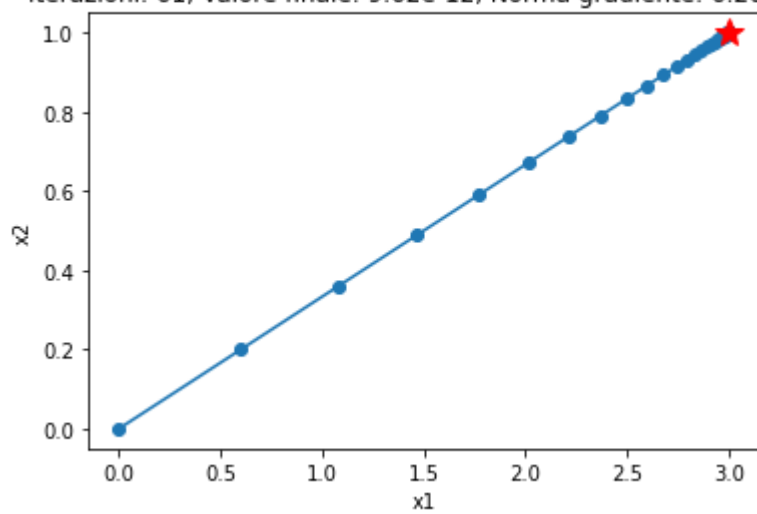
Plot function $f(x_1, x_2) = (x_1 - 3)^2 + (x_2 - 1)^2$

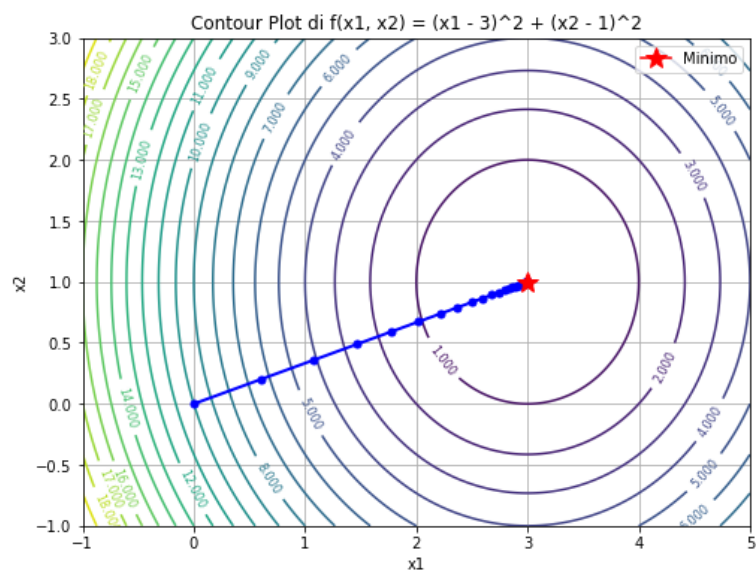


Optimization Path with Fixed α
 Iterazioni: 99, Valore finale: 1.76e-01, Norma gradiente: 8.39e-01

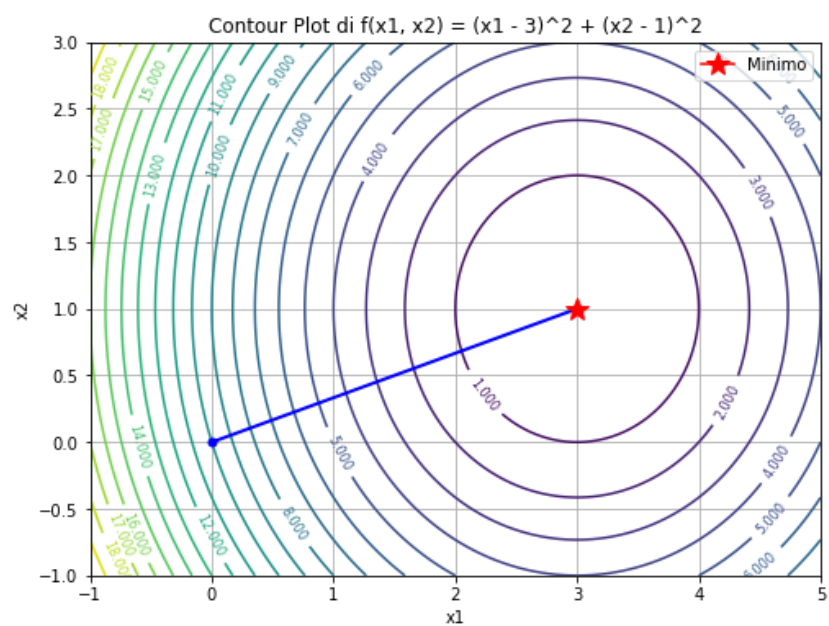
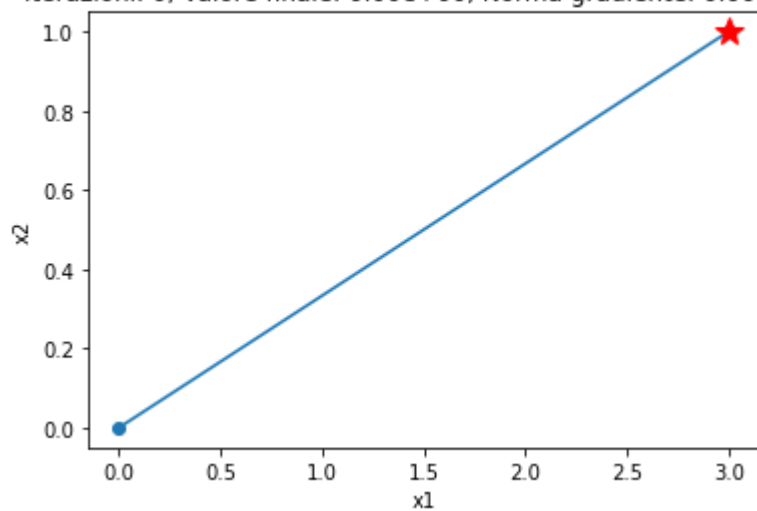


Optimization Path with Fixed α
 Iterazioni: 61, Valore finale: 9.62e-12, Norma gradiente: 6.20e-06

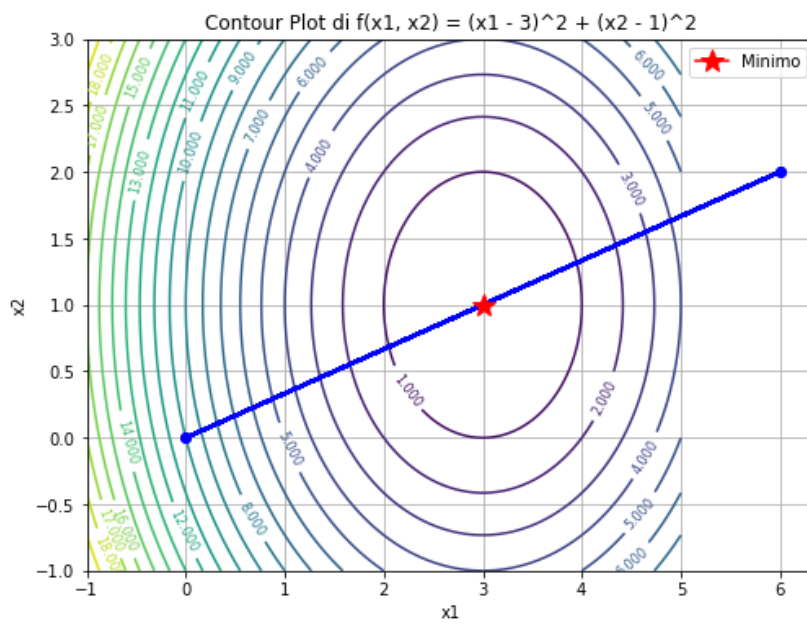
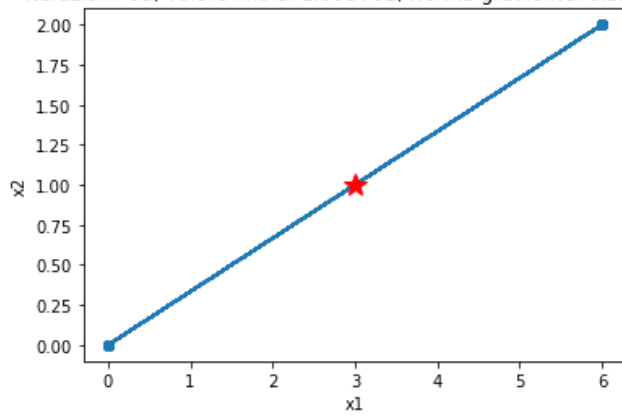




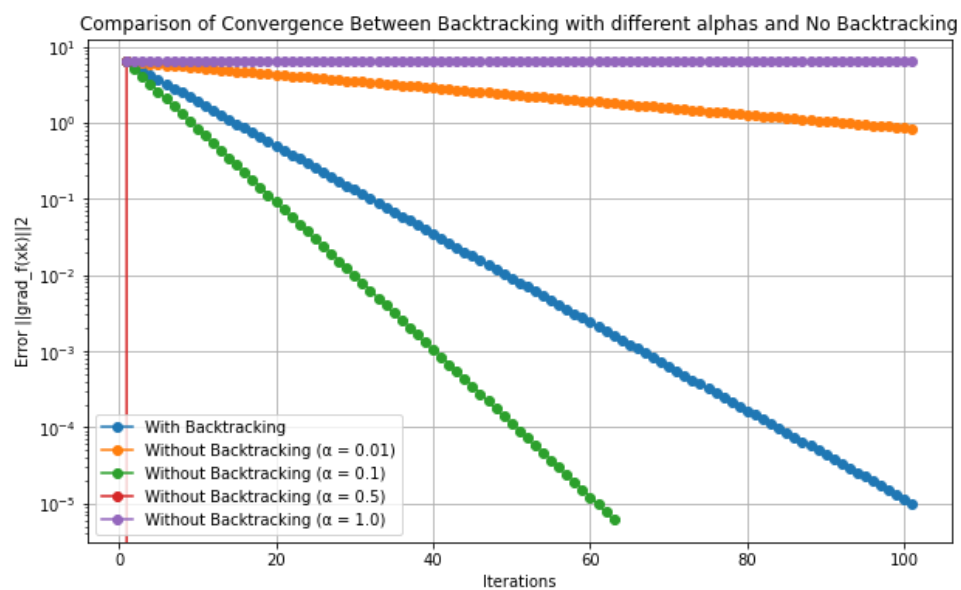
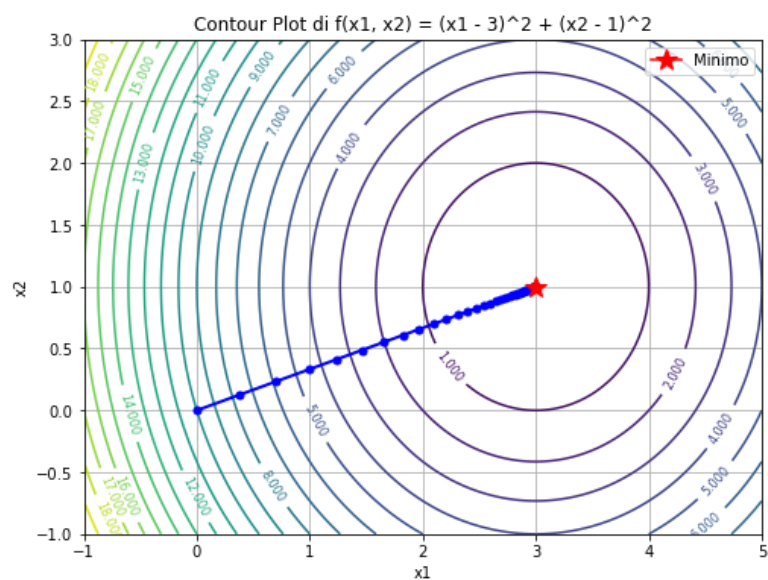
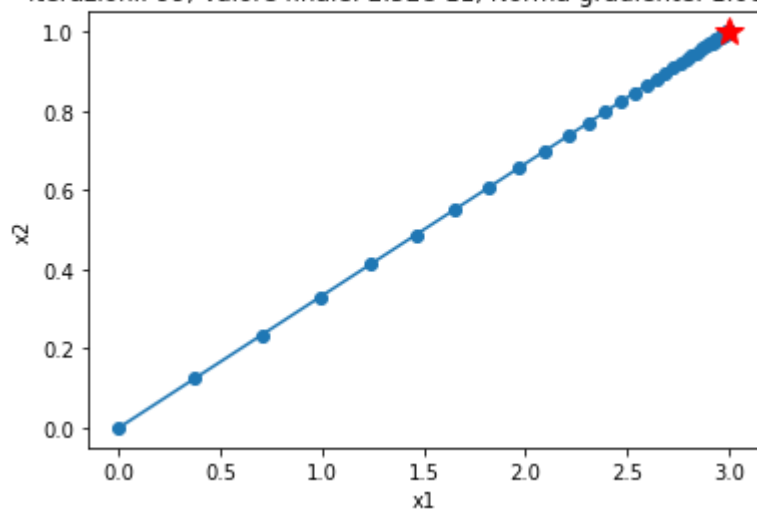
Optimization Path with Fixed α
 Iterazioni: 0, Valore finale: 0.00e+00, Norma gradiente: 0.00e+00



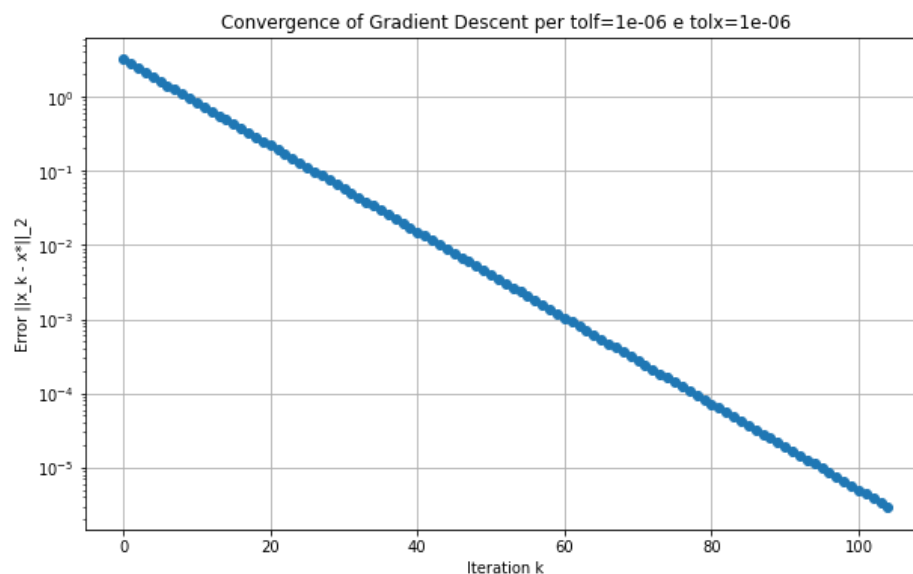
Optimization Path with Fixed α
Iterazioni: 99, Valore finale: 1.00e+01, Norma gradiente: 6.32e+00

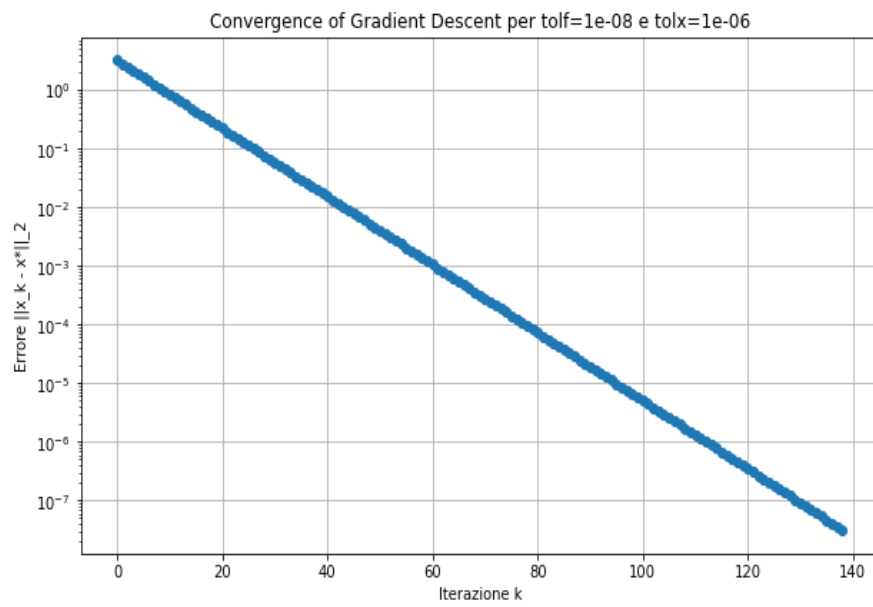
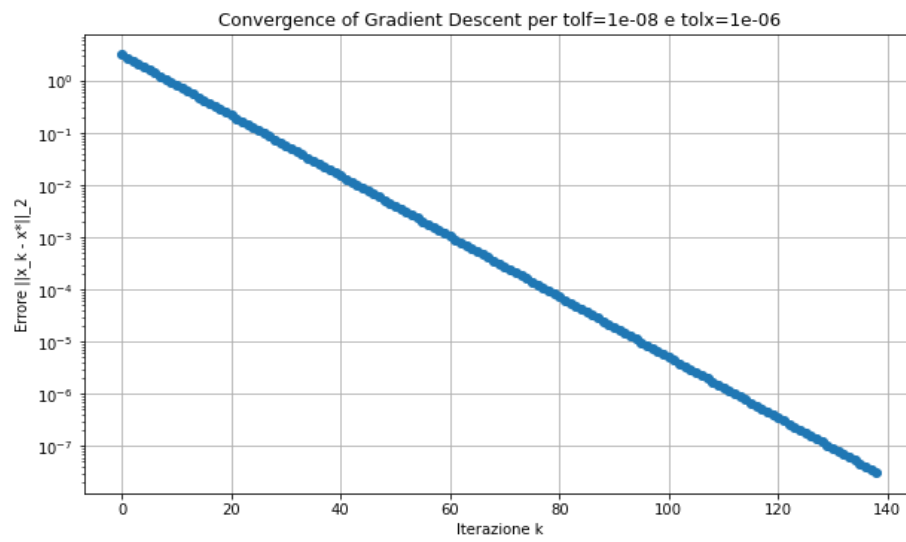


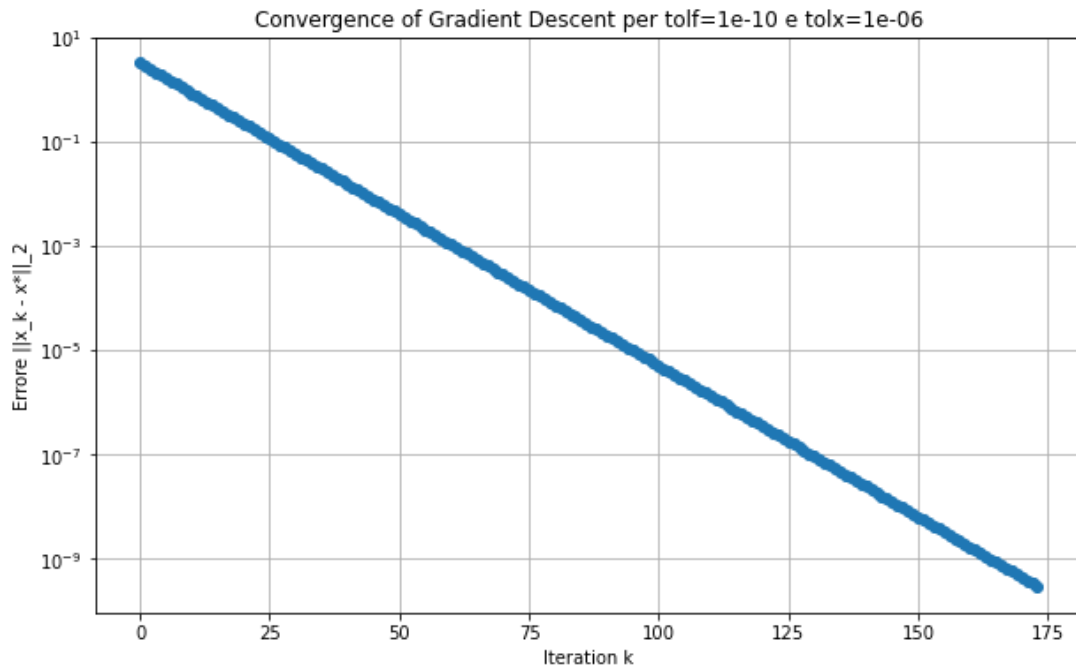
Optimization path with α backtracking
 Iterazioni: 99, Valore finale: 2.52e-11, Norma gradiente: 1.00e-05



tolf: 1e-06, tolx: 1e-06 -> Iterazioni: 103, Valore finale: 8.663133929057938e-12,
Cambiamento in x: 4.2047431734906344e-07
tolf: 1e-06, tolx: 1e-08 -> Iterazioni: 103, Valore finale: 8.663133929057938e-12,
Cambiamento in x: 4.2047431734906344e-07
tolf: 1e-06, tolx: 1e-10 -> Iterazioni: 103, Valore finale: 8.663133929057938e-12,
Cambiamento in x: 4.2047431734906344e-07
tolf: 1e-08, tolx: 1e-06 -> Iterazioni: 137, Valore finale: 9.867852111454736e-16,
Cambiamento in x: 4.487591330278101e-09
tolf: 1e-08, tolx: 1e-08 -> Iterazioni: 137, Valore finale: 9.867852111454736e-16,
Cambiamento in x: 4.487591330278101e-09
tolf: 1e-08, tolx: 1e-10 -> Iterazioni: 137, Valore finale: 9.867852111454736e-16,
Cambiamento in x: 4.487591330278101e-09
tolf: 1e-10, tolx: 1e-06 -> Iterazioni: 172, Valore finale: 8.605713131993931e-20,
Cambiamento in x: 4.190808183950403e-11
tolf: 1e-10, tolx: 1e-08 -> Iterazioni: 172, Valore finale: 8.605713131993931e-20,
Cambiamento in x: 4.190808183950403e-11
tolf: 1e-10, tolx: 1e-10 -> Iterazioni: 172, Valore finale: 8.605713131993931e-20,
Cambiamento in x: 4.190808183950403e-11







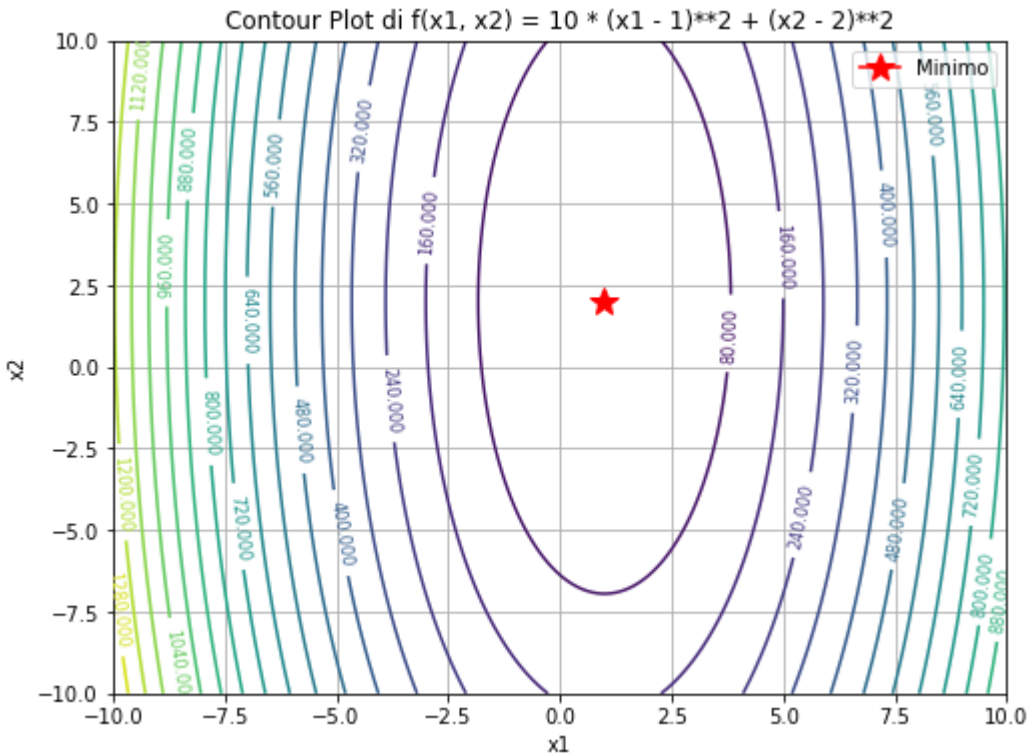
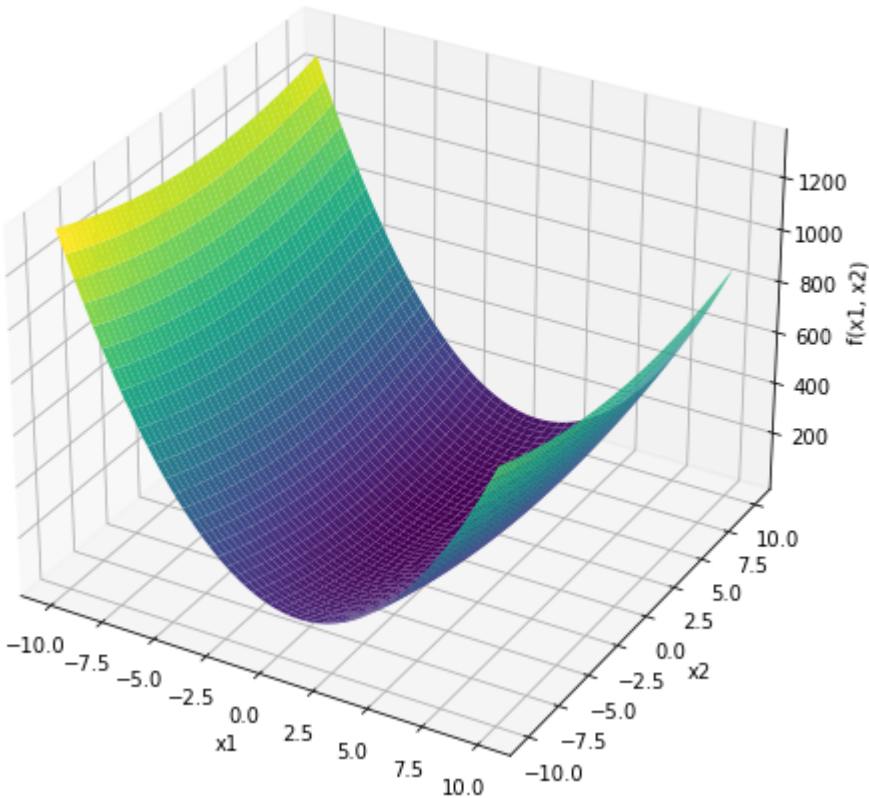
In conducting an analysis of the gradient descent algorithm's performance with varying tolerances for the objective function tolf and decision variables tolx , it was observed that the convergence of the algorithm is predominantly influenced by the tolerance on the objective function (tolf).

Specifically, as tolf is tightened (lowered from $1\text{e-}6$ to $1\text{e-}10$), the precision of the solution notably increases, as evidenced by a significant reduction in the final objective function value.

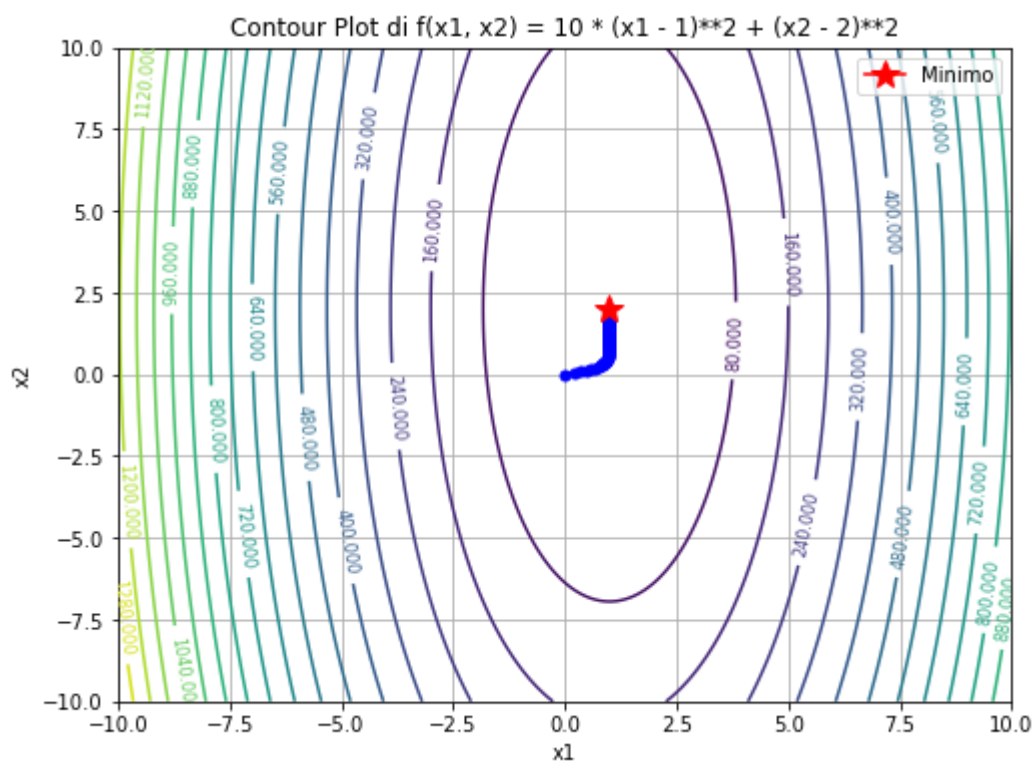
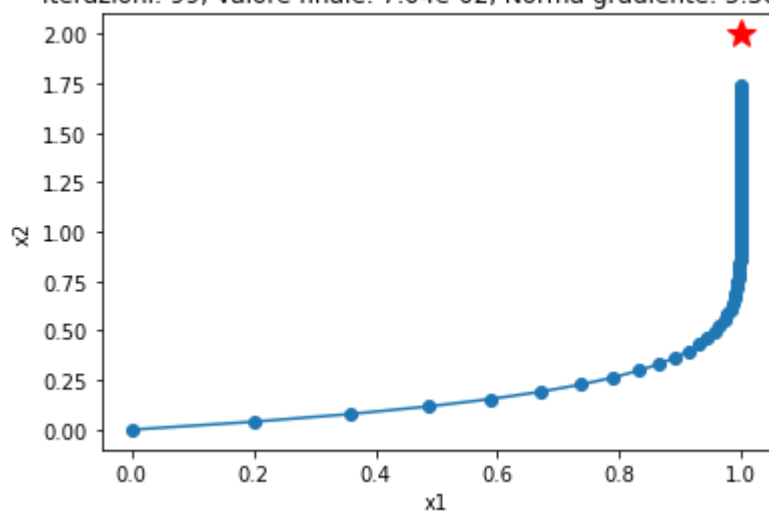
However, this increase in precision comes with a corresponding rise in the number of iterations, highlighting a classic trade-off between computational cost and solution accuracy.

Notably, changes in tolx did not materially affect the number of iterations or the final value of the objective function, suggesting that the convergence criterion based on the gradient norm (tolf) is the primary determinant in the stopping condition of the algorithm under the tested scenarios.

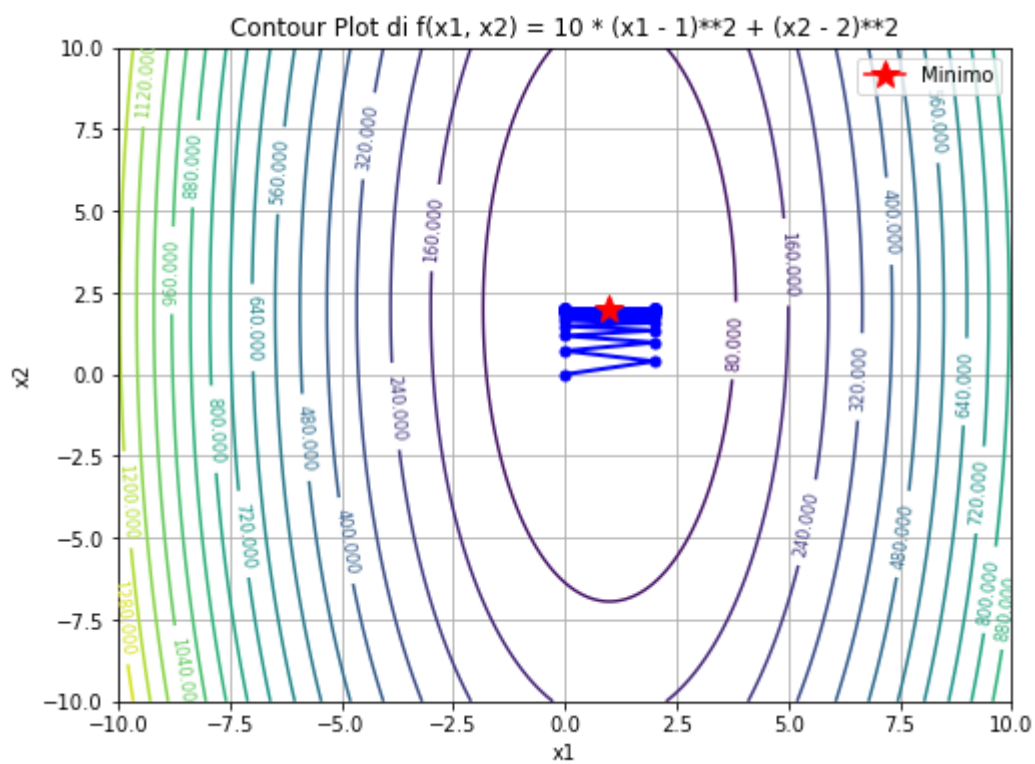
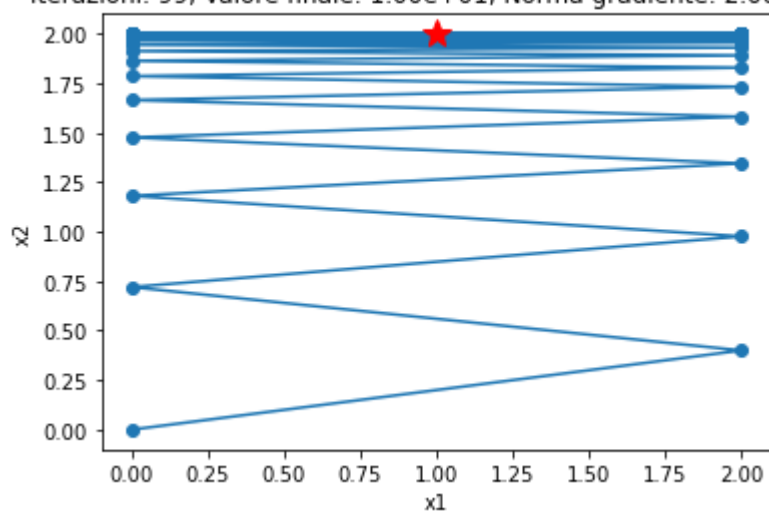
Plot of the function $f(x_1, x_2) = 10 * (x_1 - 1)**2 + (x_2 - 2)**2$



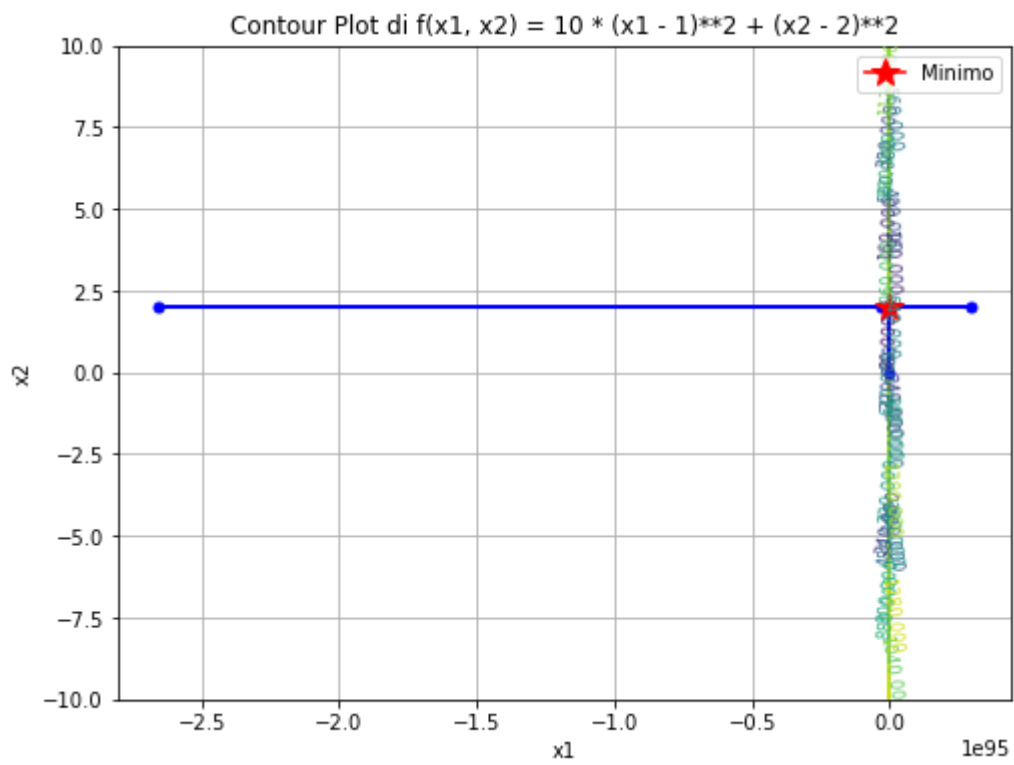
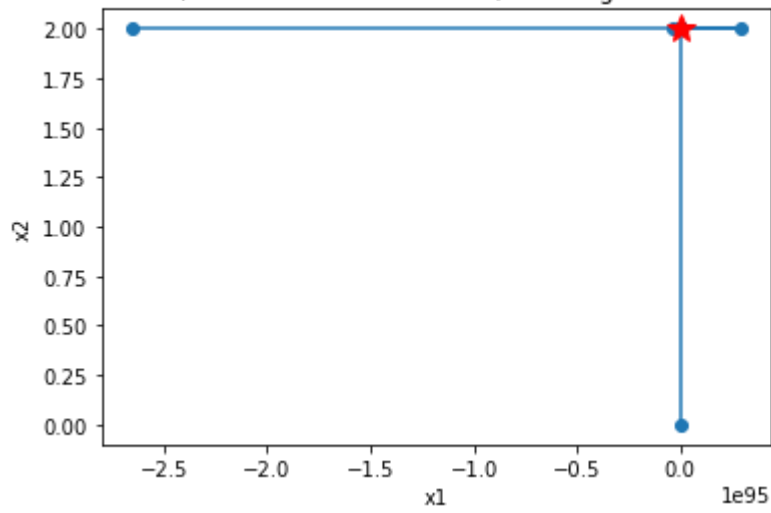
Percorso di ottimizzazione con $\alpha = 0.01$
 Iterazioni: 99, Valore finale: 7.04e-02, Norma gradiente: 5.30e-01



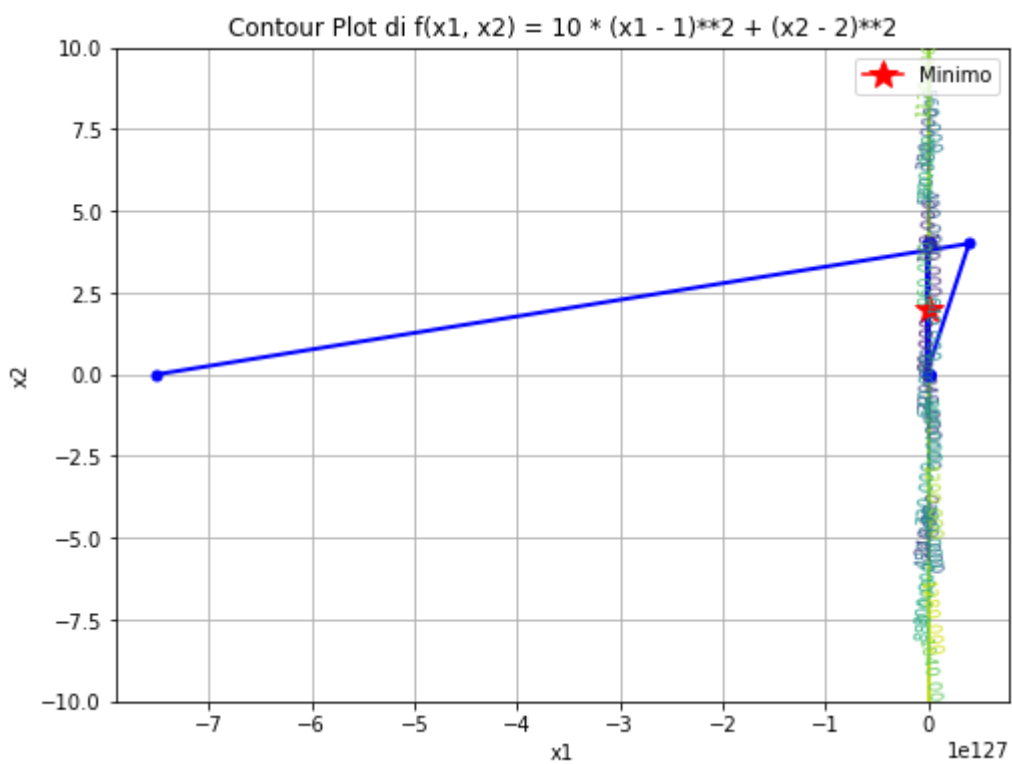
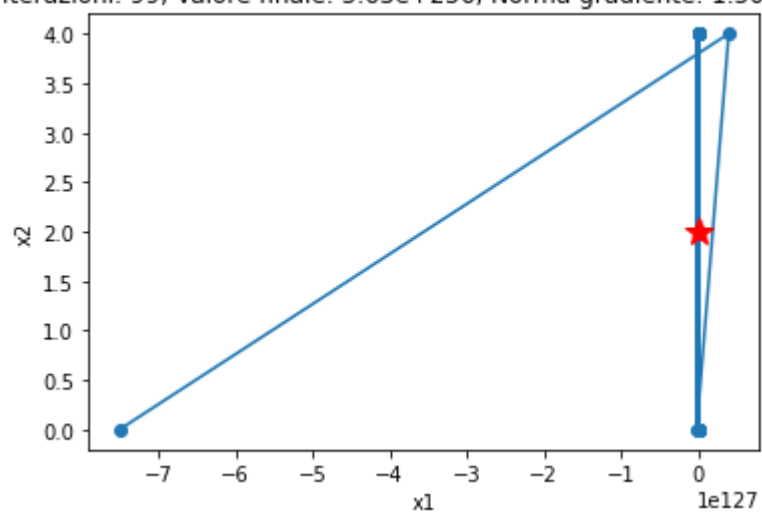
Percorso di ottimizzazione con $\alpha = 0.1$
 Iterazioni: 99, Valore finale: 1.00e+01, Norma gradiente: 2.00e+01



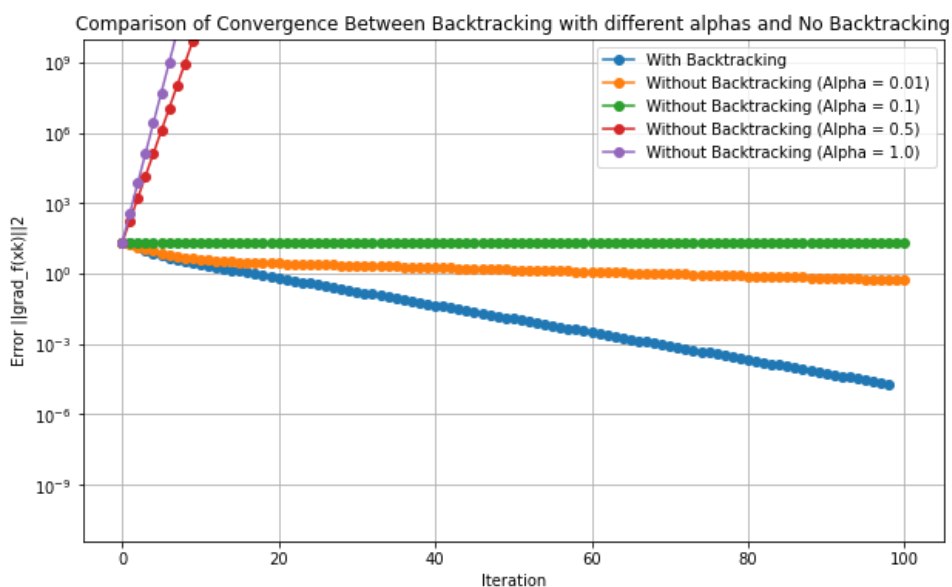
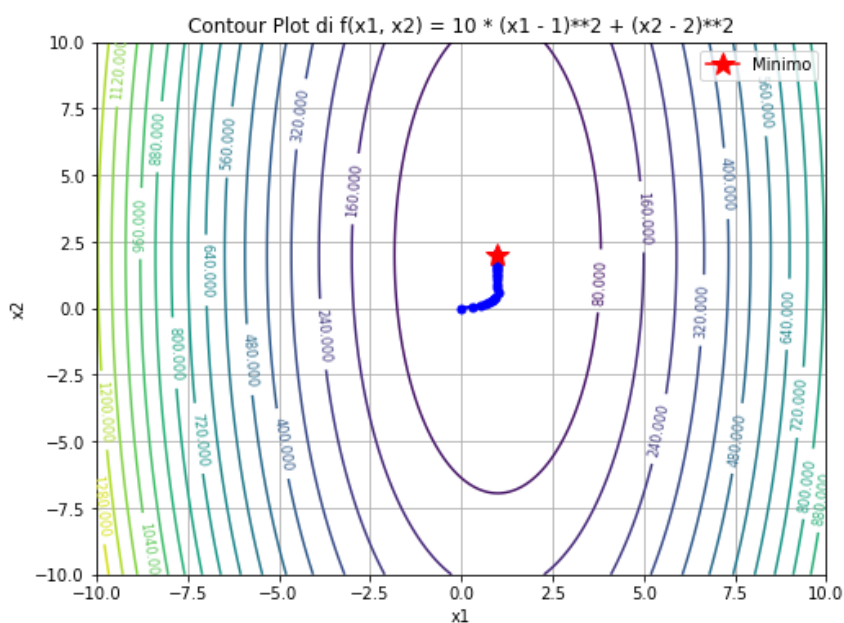
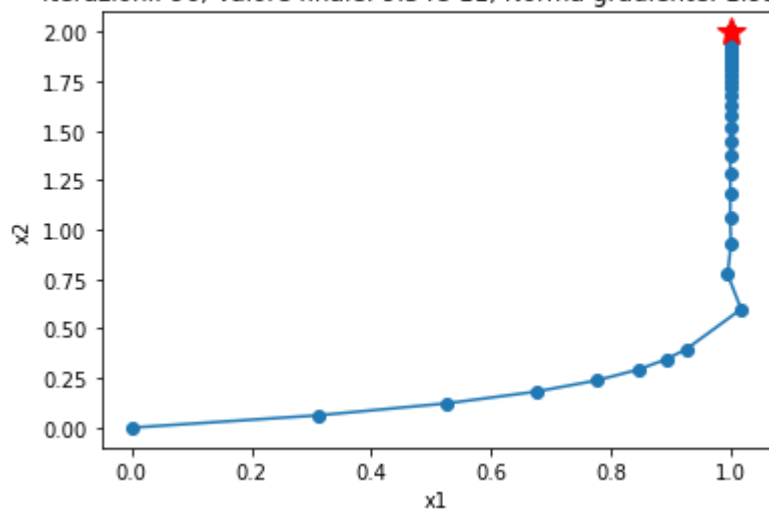
Percorso di ottimizzazione con $\alpha = 0.5$
 Iterazioni: 99, Valore finale: $7.06e+191$, Norma gradiente: $5.31e+96$



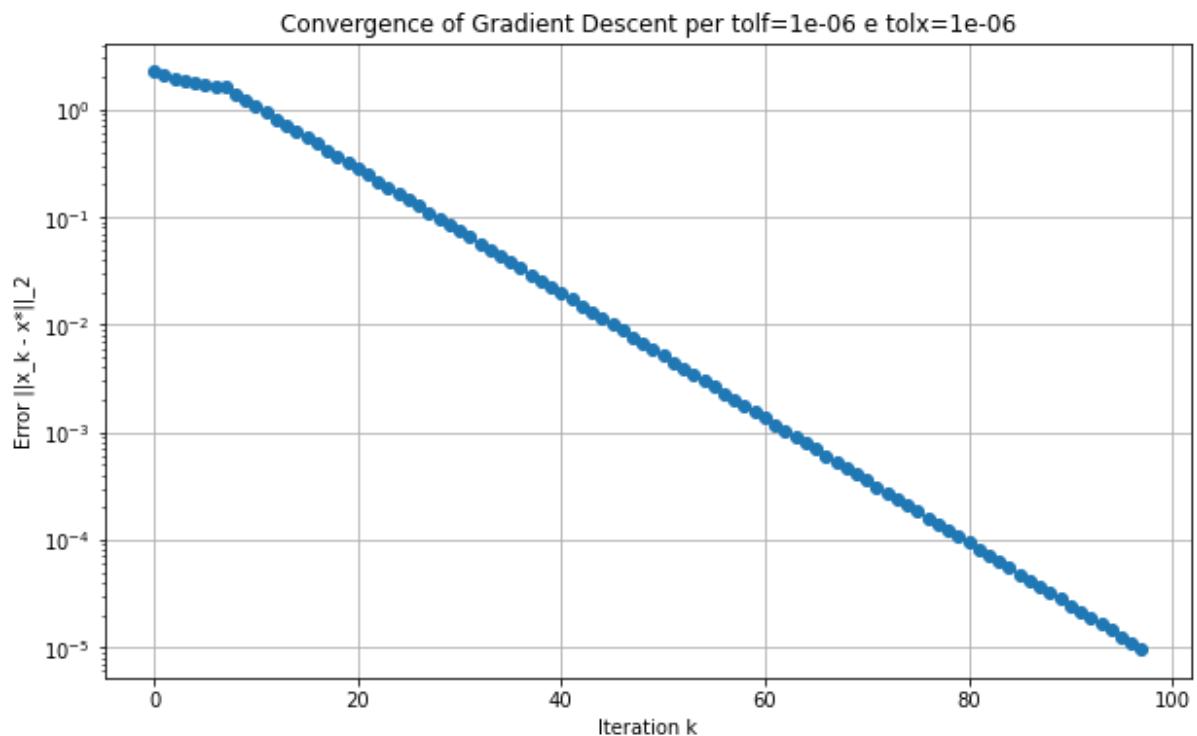
Percorso di ottimizzazione con $\alpha = 1.0$
 Iterazioni: 99, Valore finale: 5.63e+256, Norma gradiente: 1.50e+129

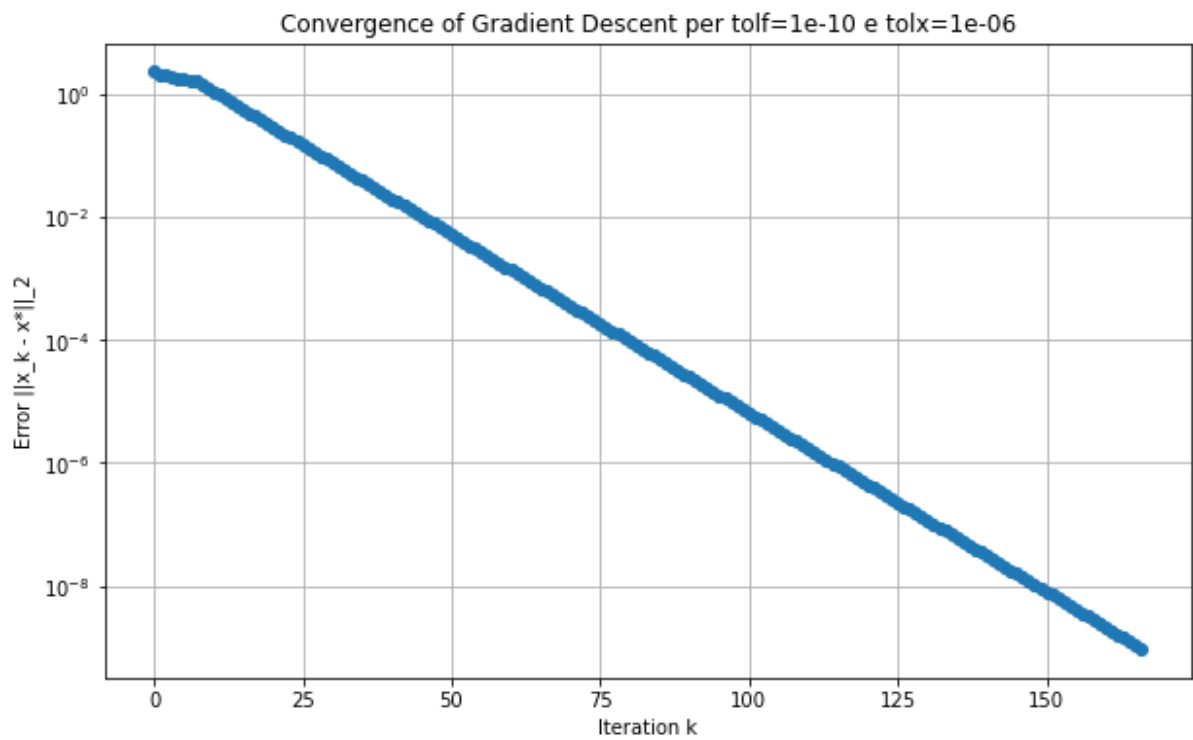
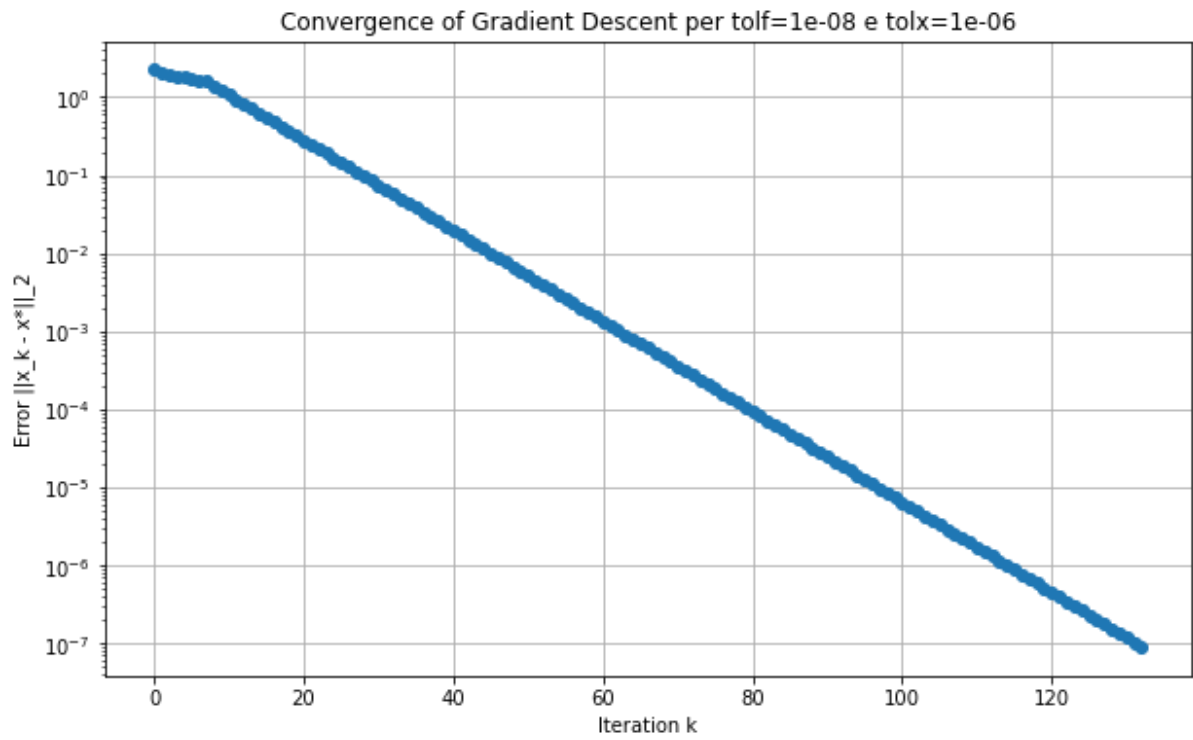


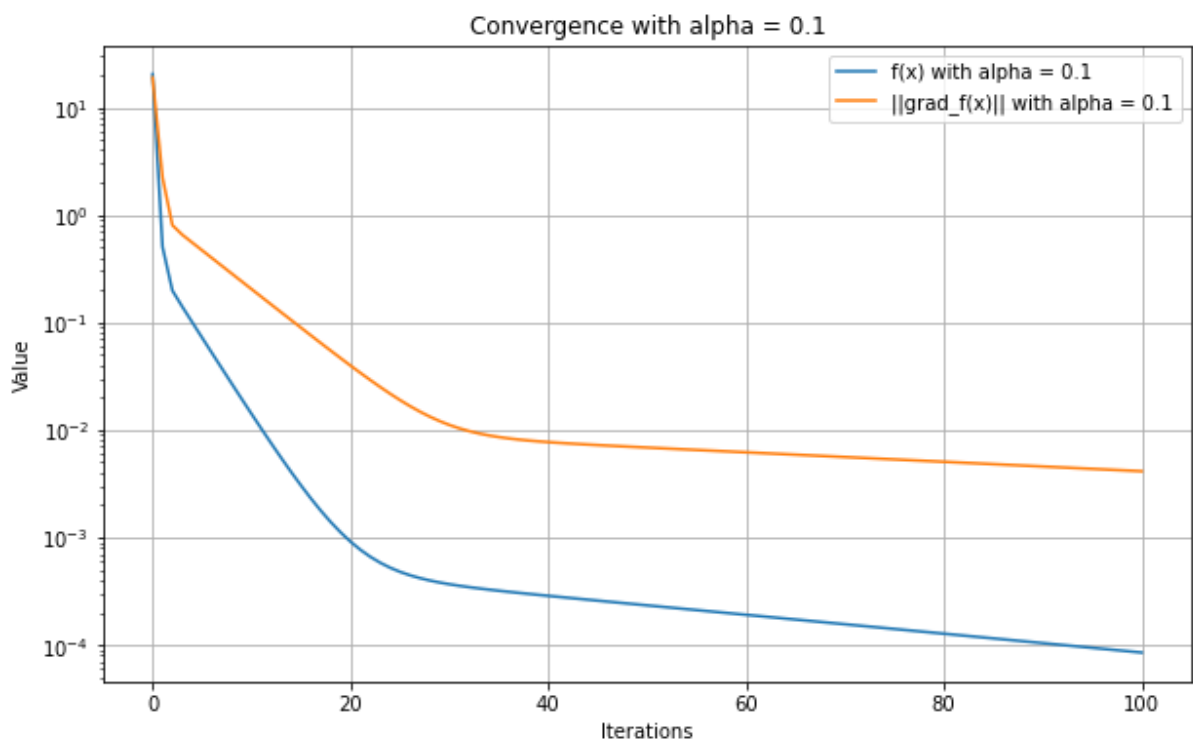
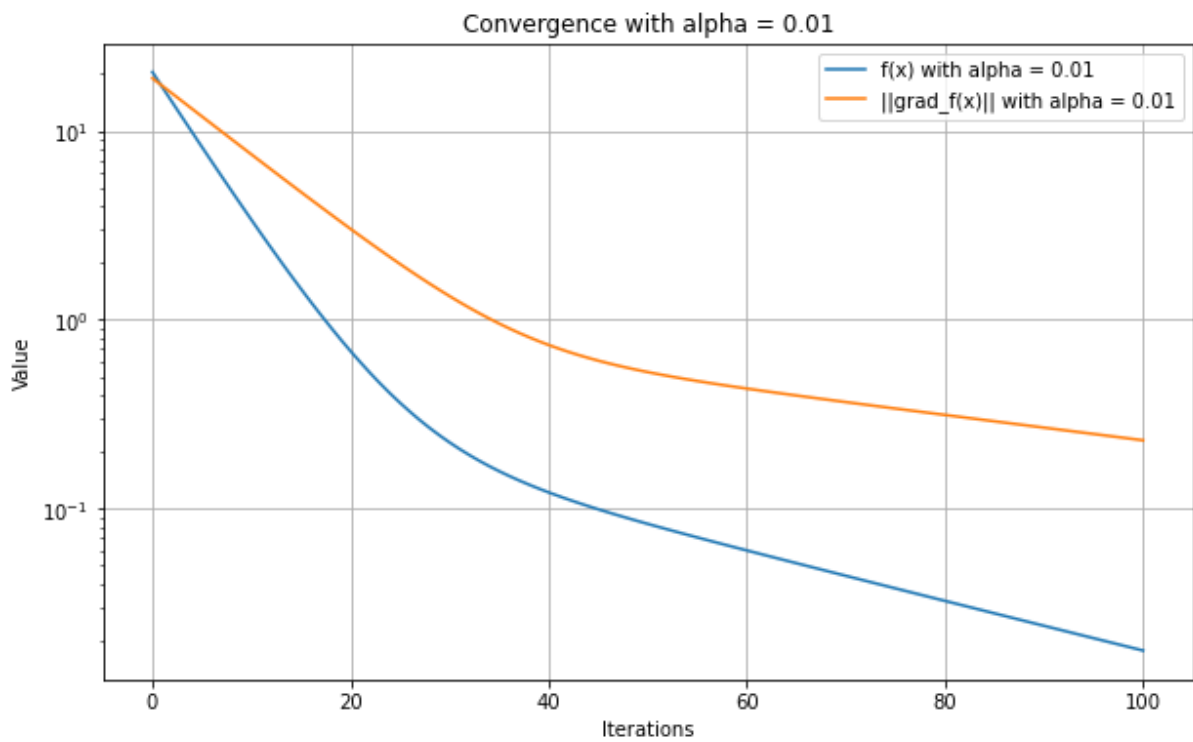
Percorso di ottimizzazione con α backtracking
 Iterazioni: 96, Valore finale: $9.34\text{e-}11$, Norma gradiente: $1.93\text{e-}05$

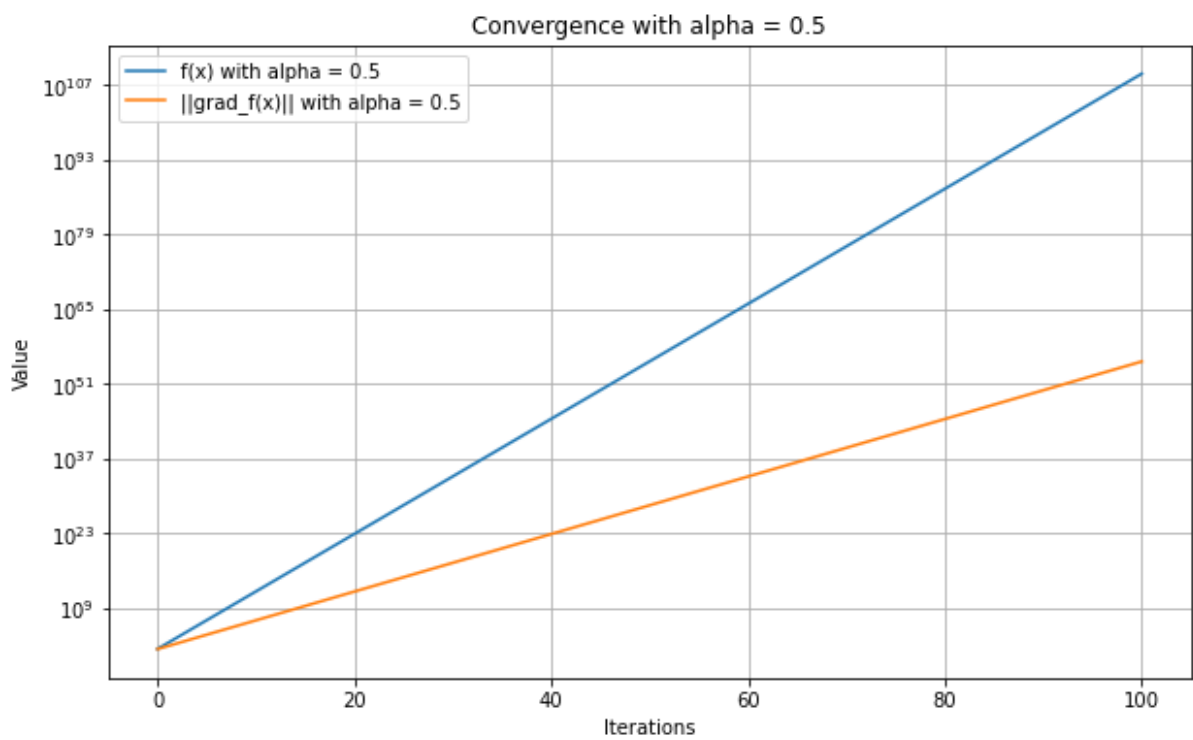
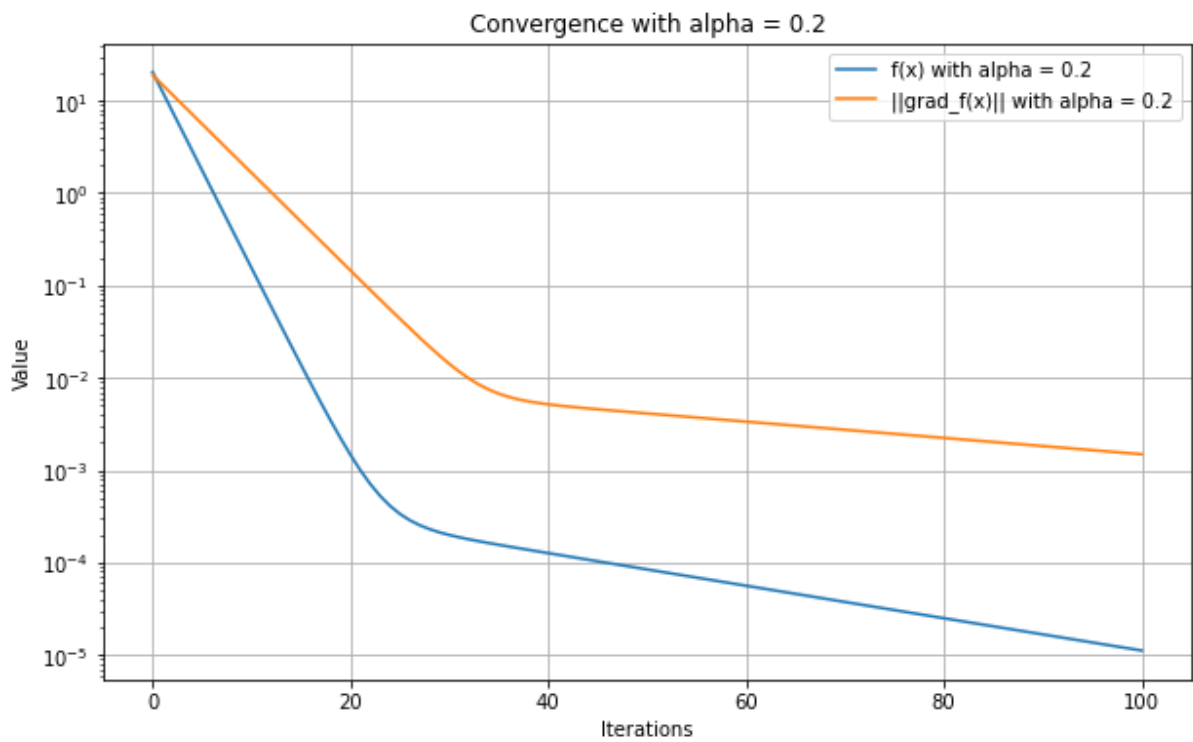


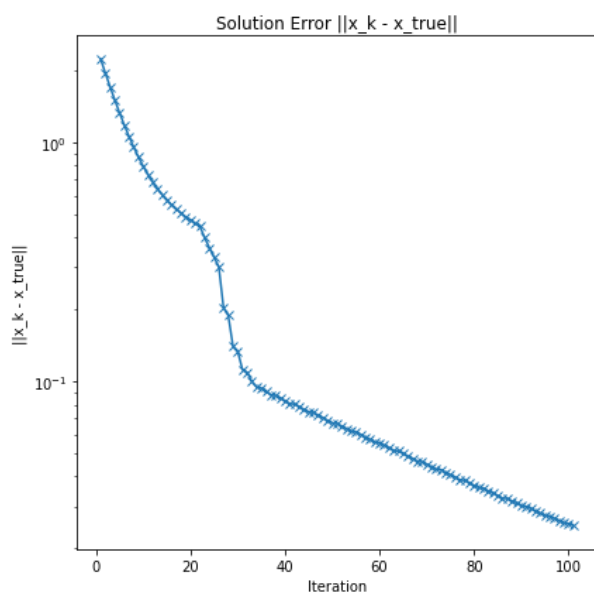
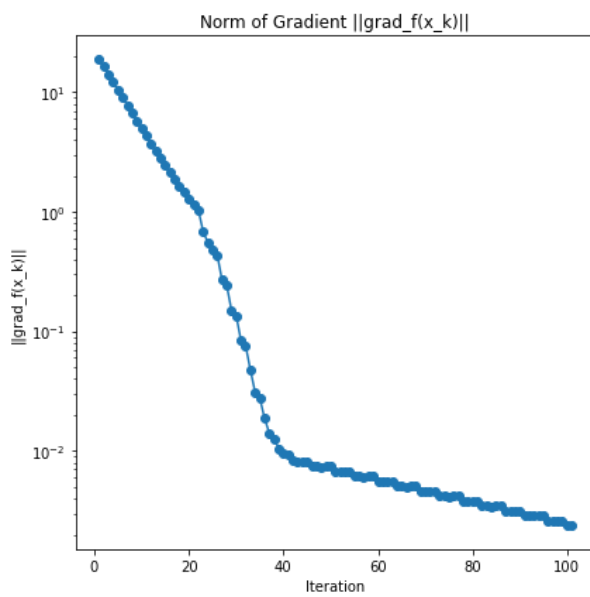
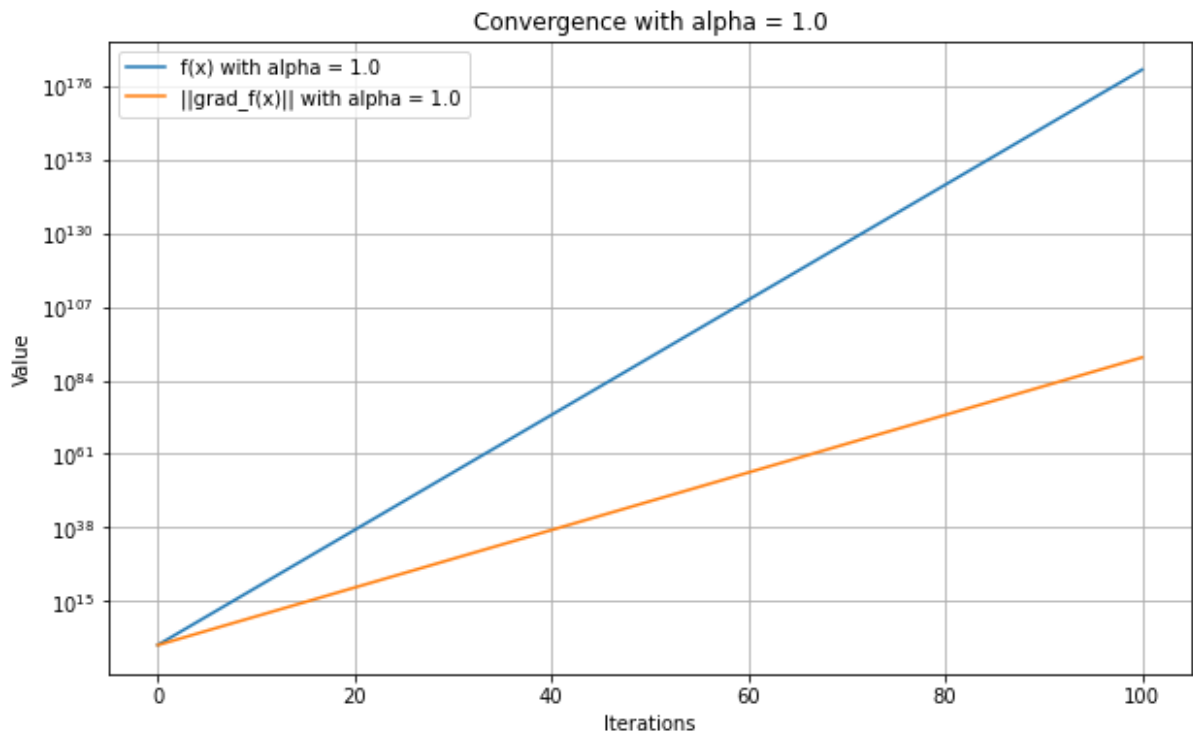
tolf: 1e-06, tolx: 1e-06 -> Iterazioni: 96, Valore finale: 9.342724716663181e-11,
 Cambiamento in x: 1.3808253016556193e-06
 tolf: 1e-06, tolx: 1e-08 -> Iterazioni: 96, Valore finale: 9.342724716663181e-11,
 Cambiamento in x: 1.3808253016556193e-06
 tolf: 1e-06, tolx: 1e-10 -> Iterazioni: 96, Valore finale: 9.342724716663181e-11,
 Cambiamento in x: 1.3808253016556193e-06
 tolf: 1e-08, tolx: 1e-06 -> Iterazioni: 131, Valore finale: 8.147741909896392e-15,
 Cambiamento in x: 1.2894977530208962e-08
 tolf: 1e-08, tolx: 1e-08 -> Iterazioni: 131, Valore finale: 8.147741909896392e-15,
 Cambiamento in x: 1.2894977530208962e-08
 tolf: 1e-08, tolx: 1e-10 -> Iterazioni: 131, Valore finale: 8.147741909896392e-15,
 Cambiamento in x: 1.2894977530208962e-08
 tolf: 1e-10, tolx: 1e-06 -> Iterazioni: 165, Valore finale: 9.280784314323266e-19,
 Cambiamento in x: 1.376241343109541e-10
 tolf: 1e-10, tolx: 1e-08 -> Iterazioni: 165, Valore finale: 9.280784314323266e-19,
 Cambiamento in x: 1.376241343109541e-10
 tolf: 1e-10, tolx: 1e-10 -> Iterazioni: 165, Valore finale: 9.280784314323266e-19,
 Cambiamento in x: 1.376241343109541e-10





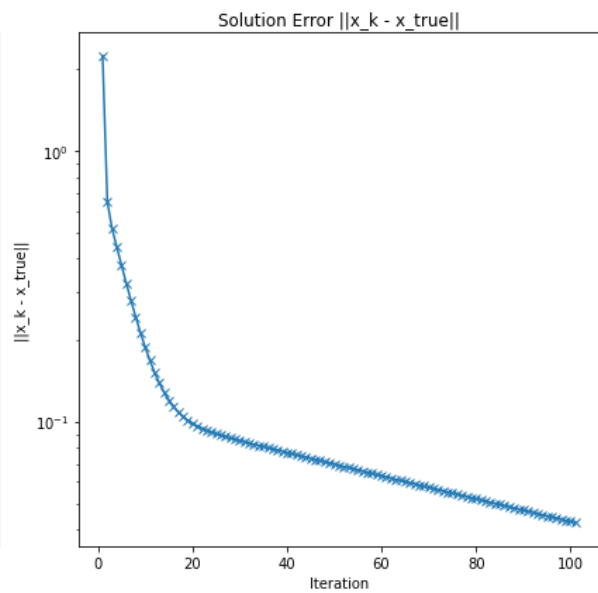
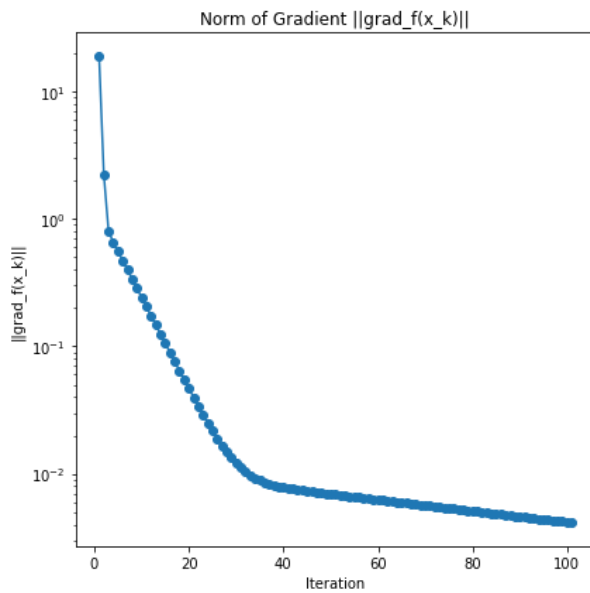
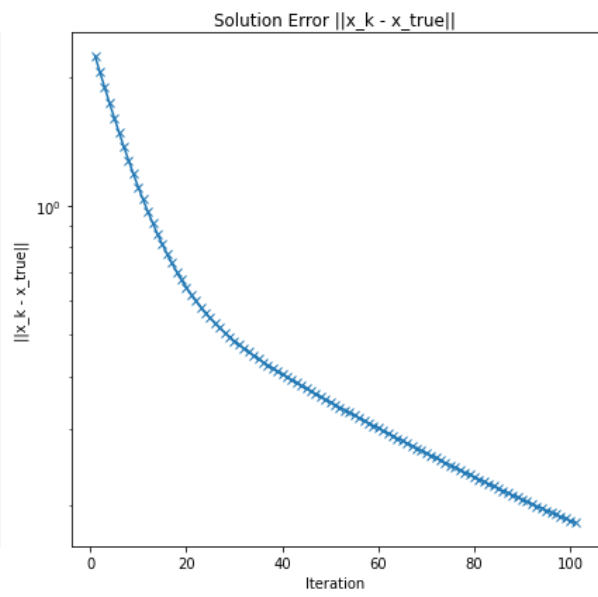
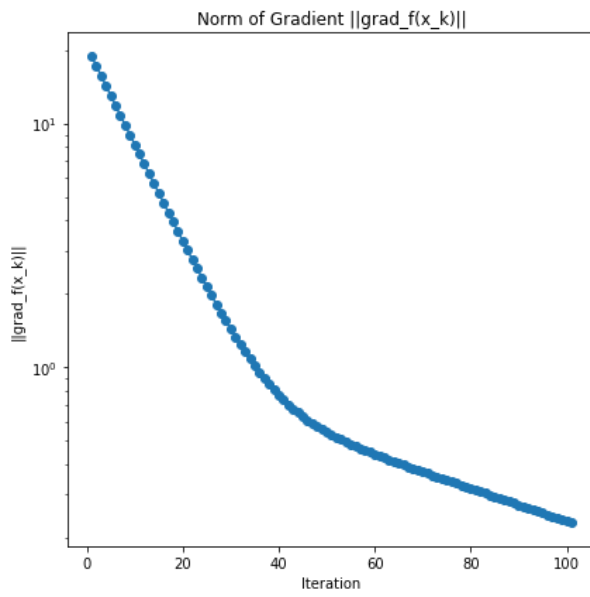


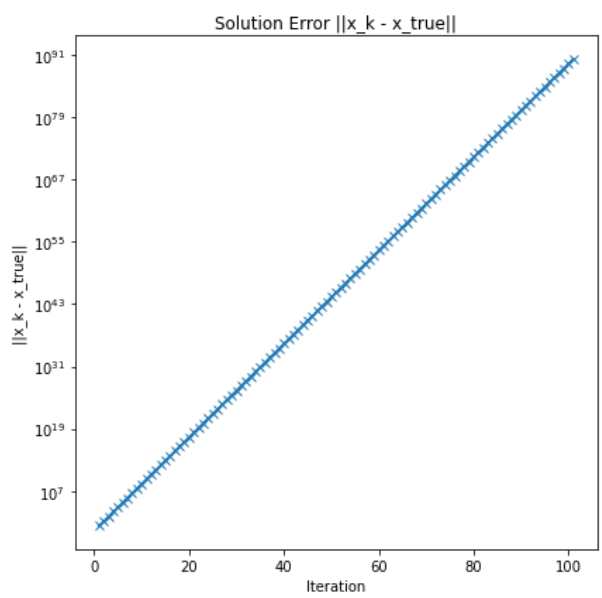
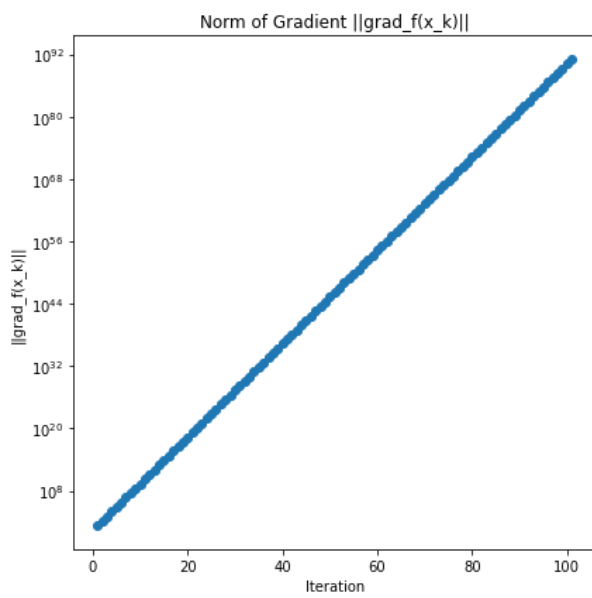
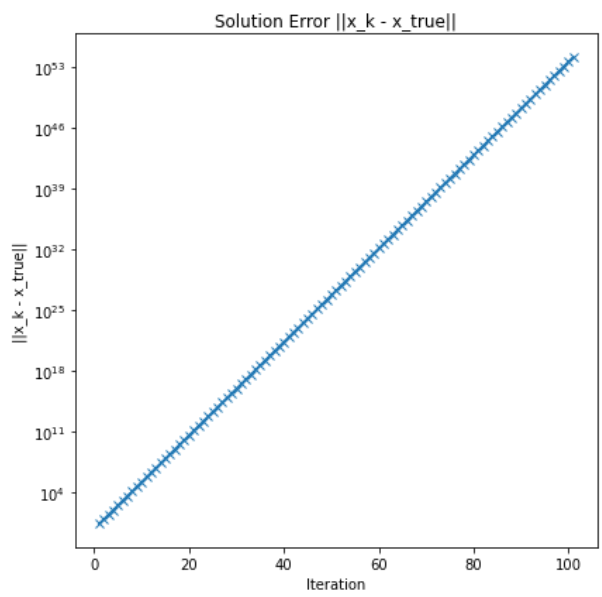
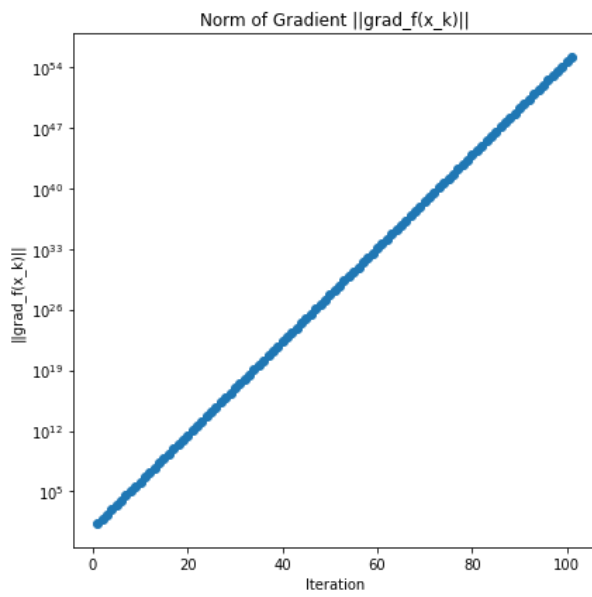


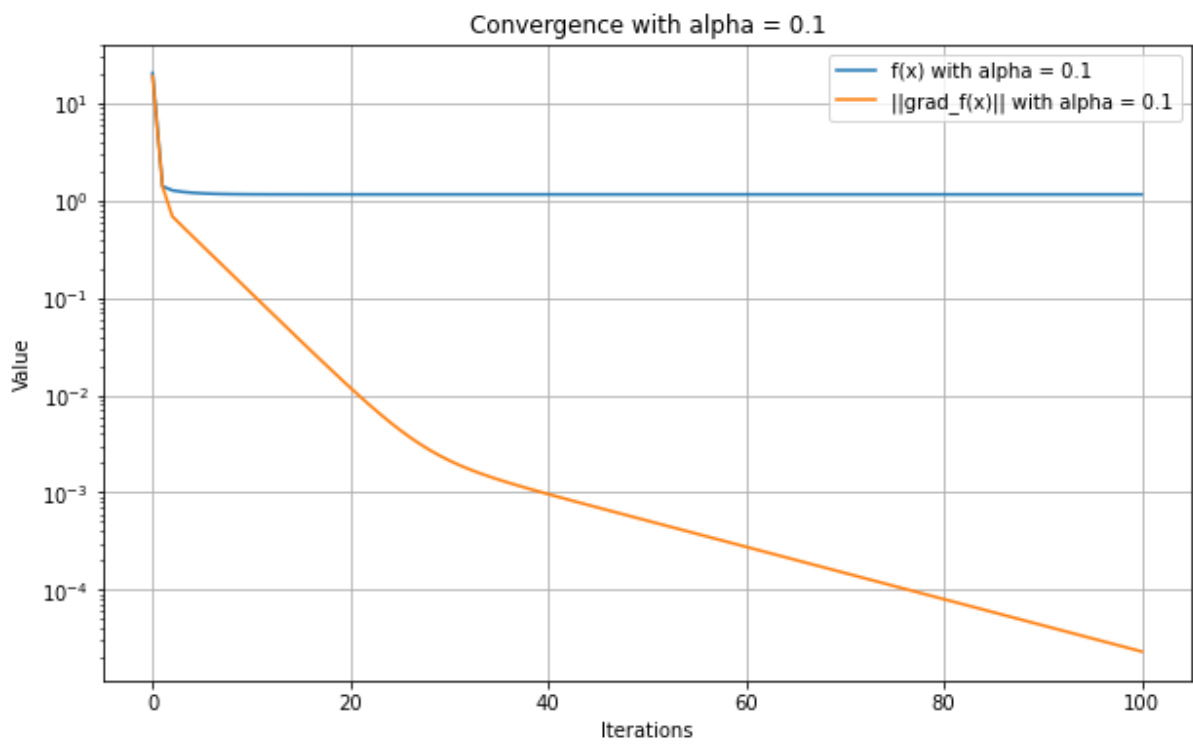
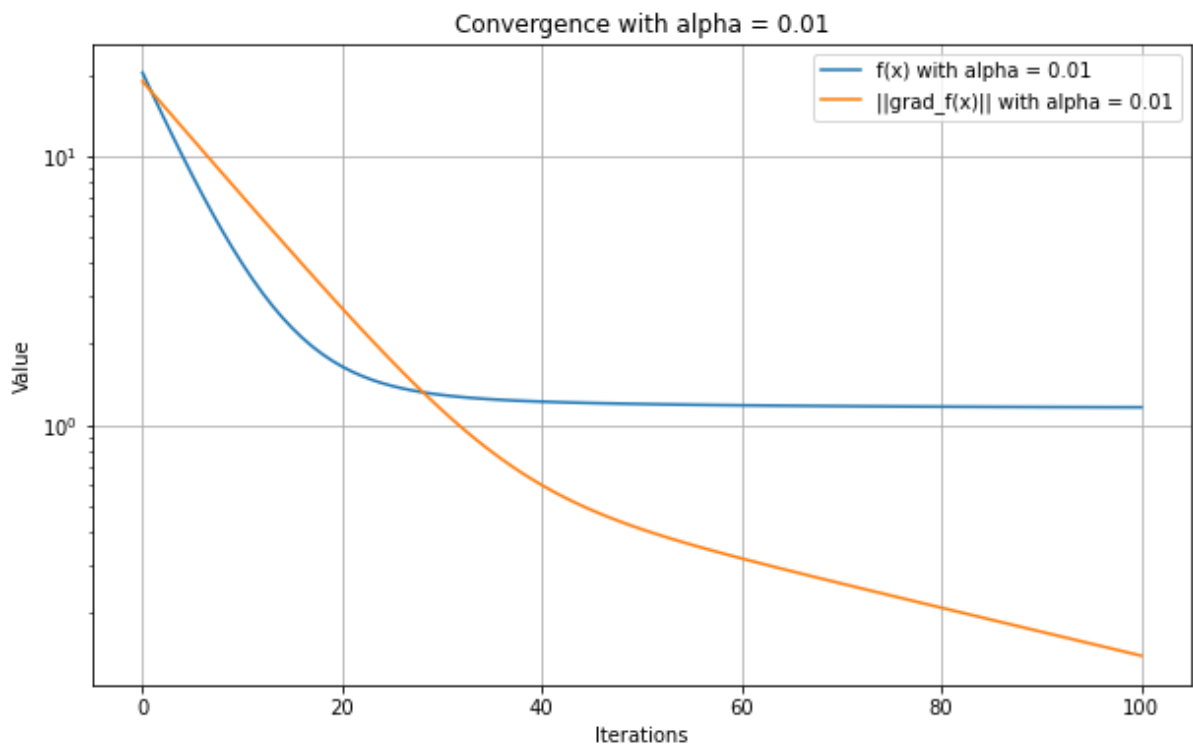


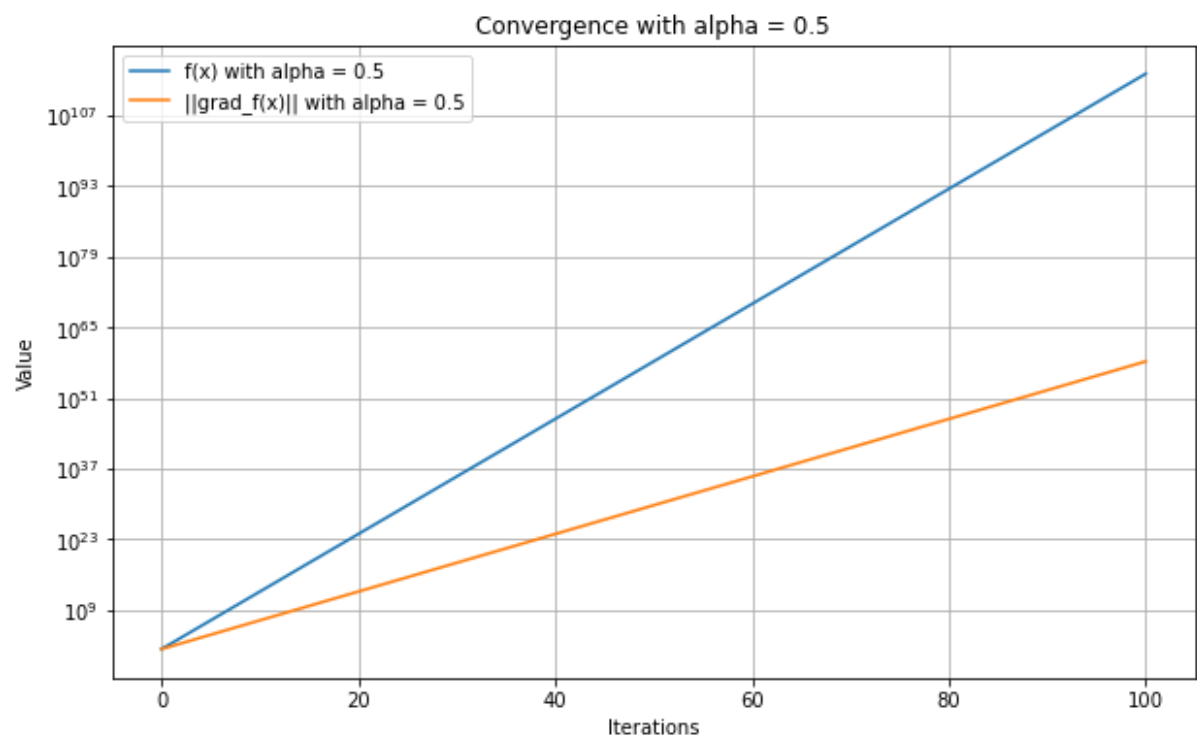
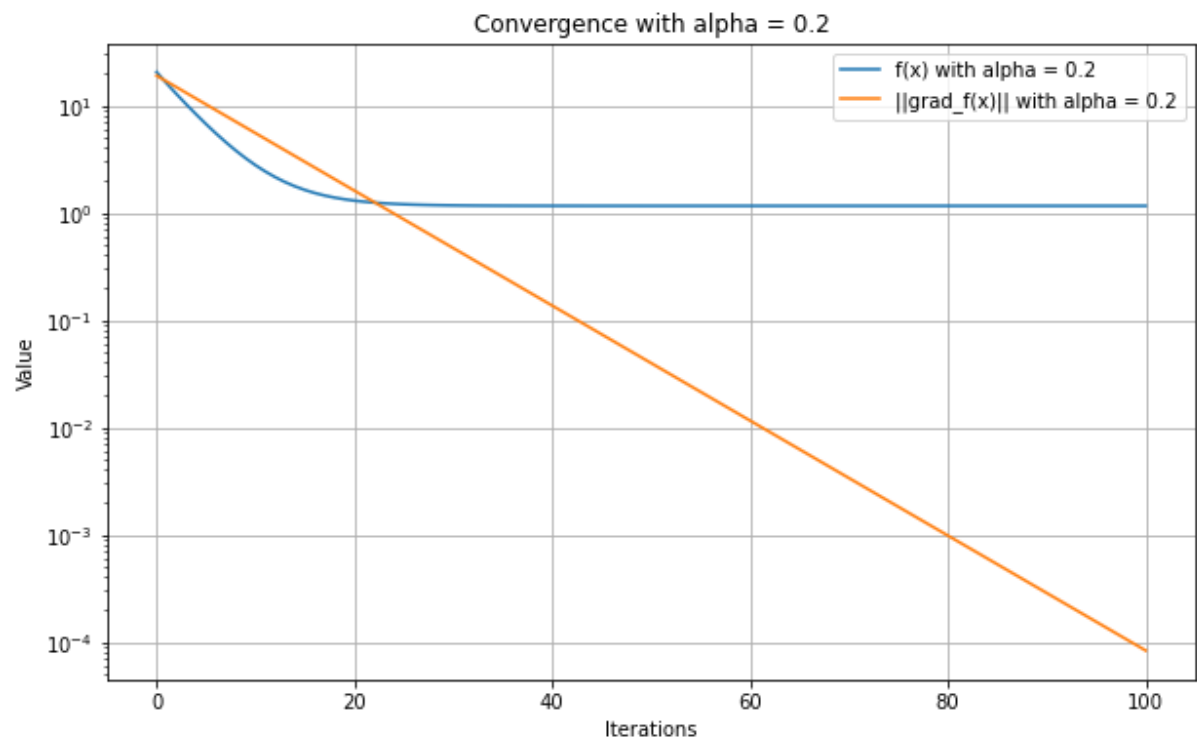
Backtracking

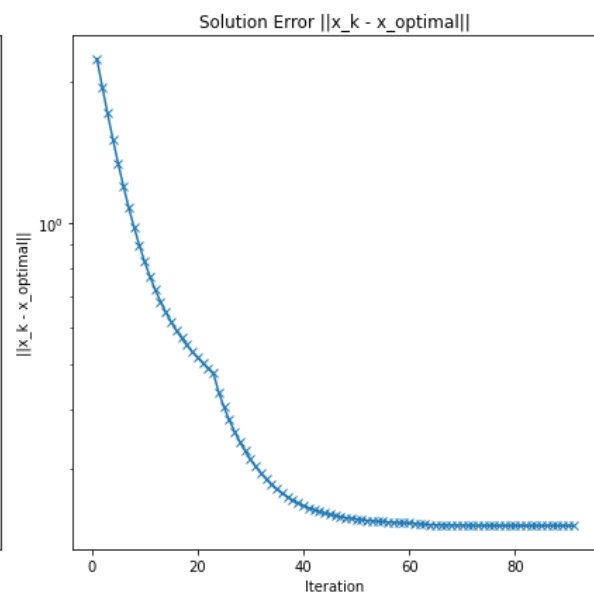
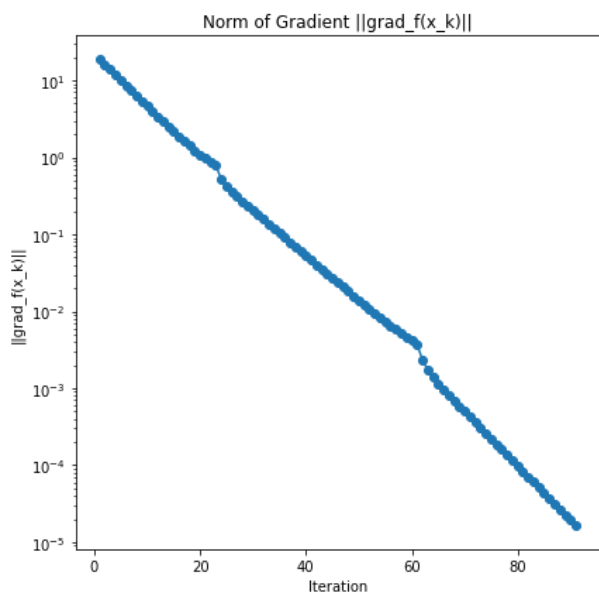
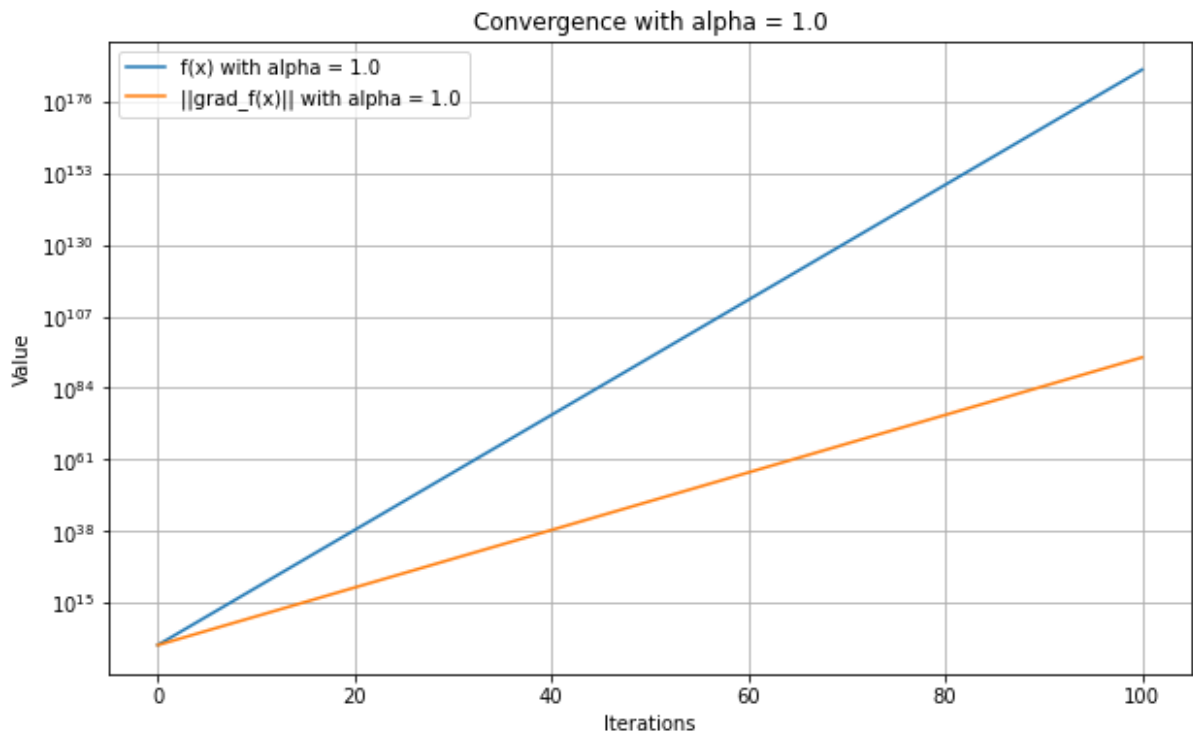
No backtracking





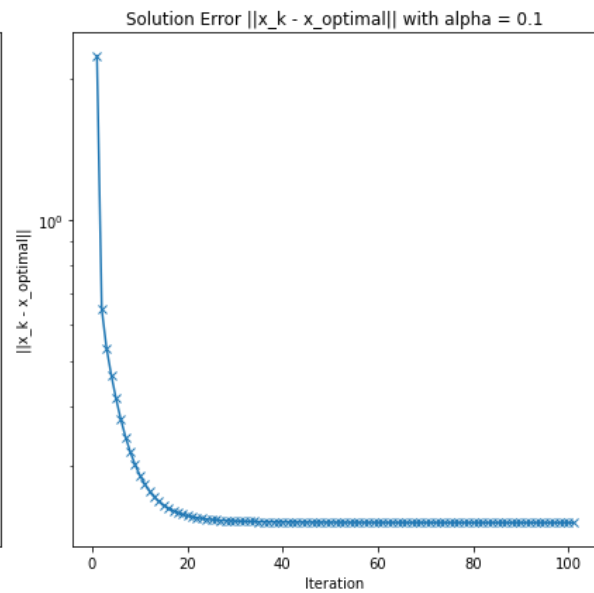
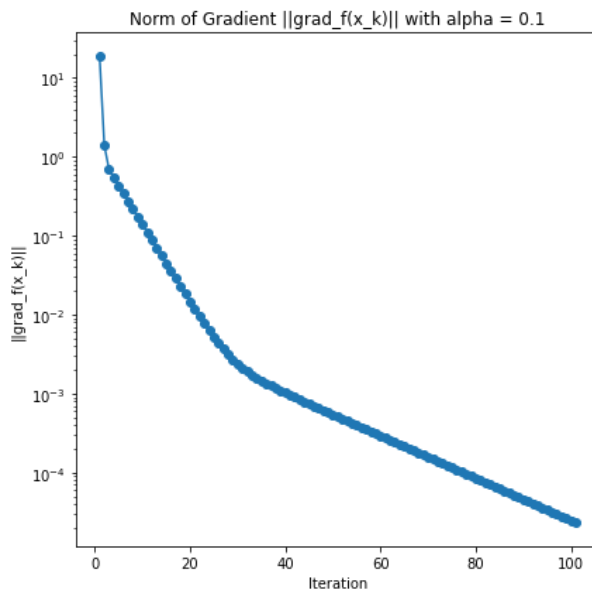
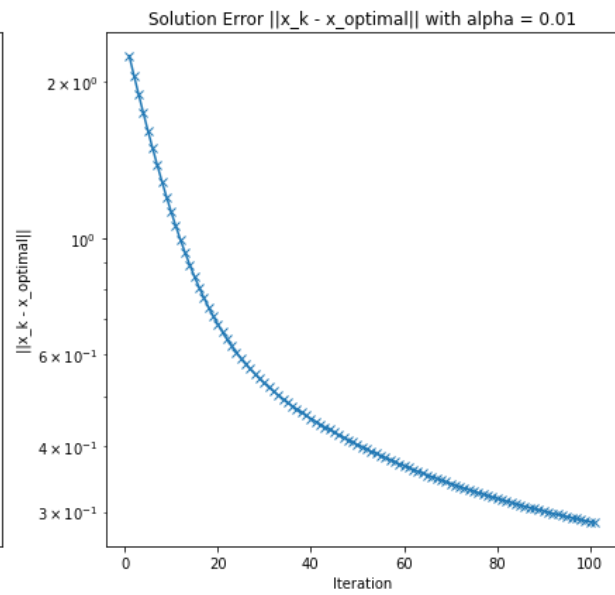
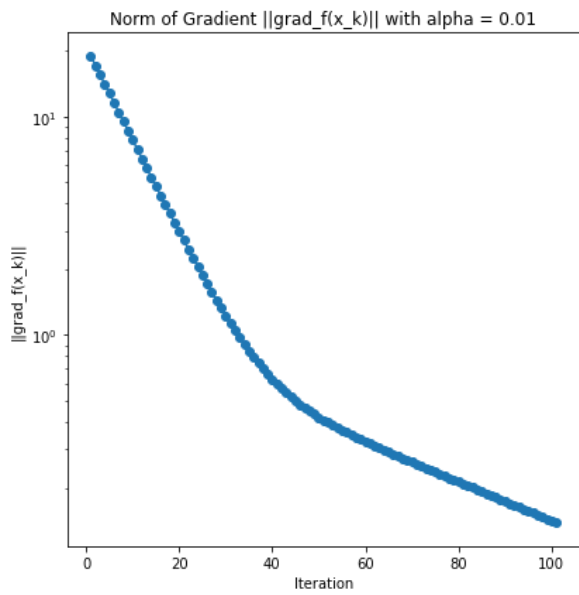


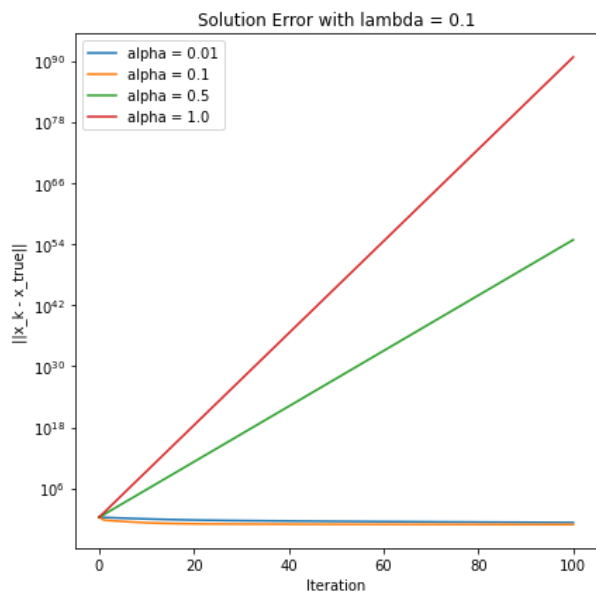
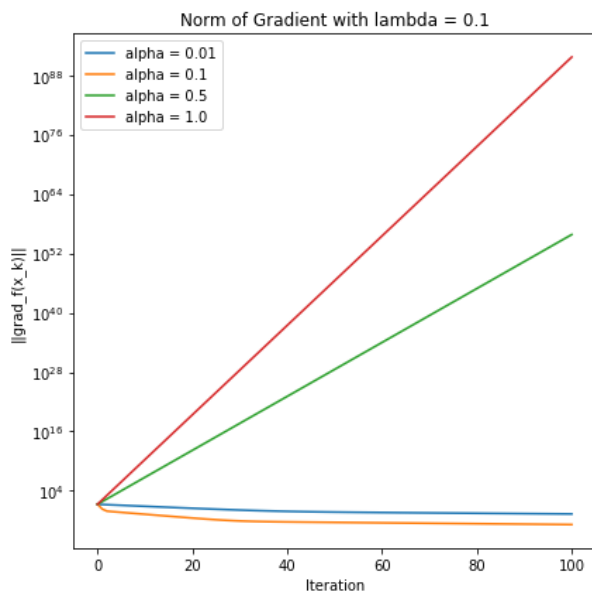
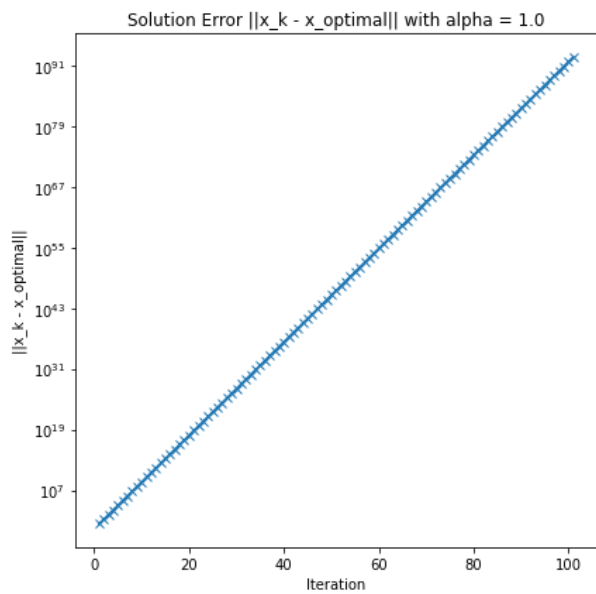
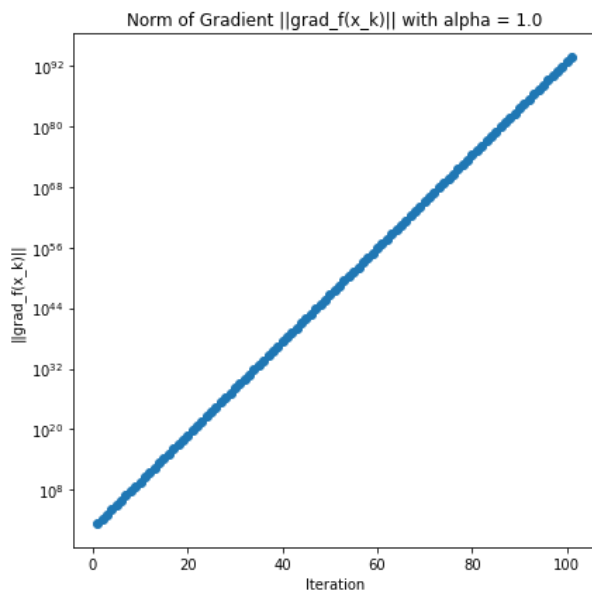
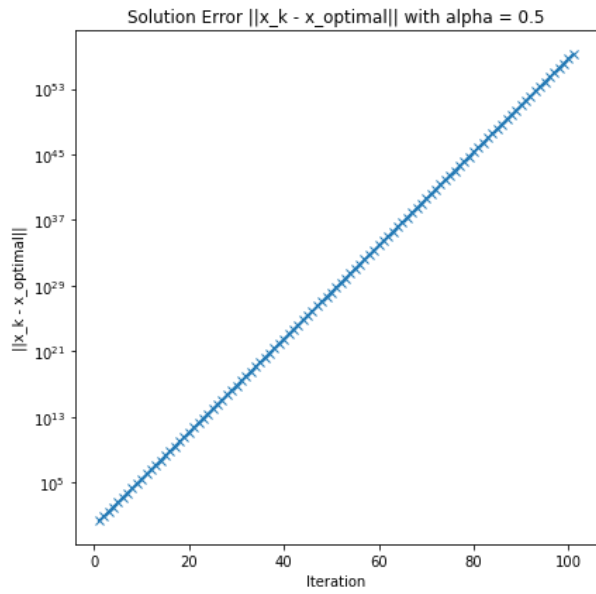
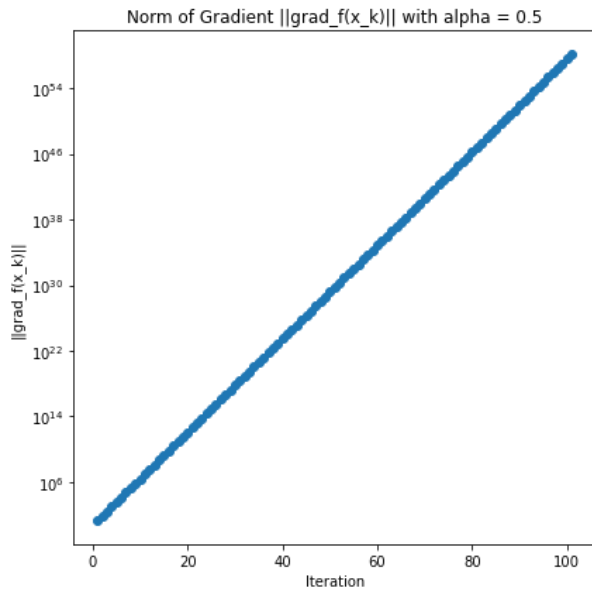


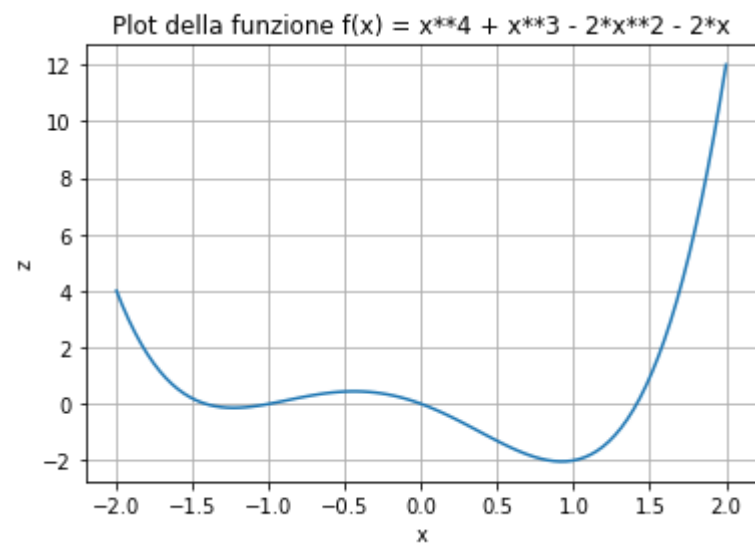
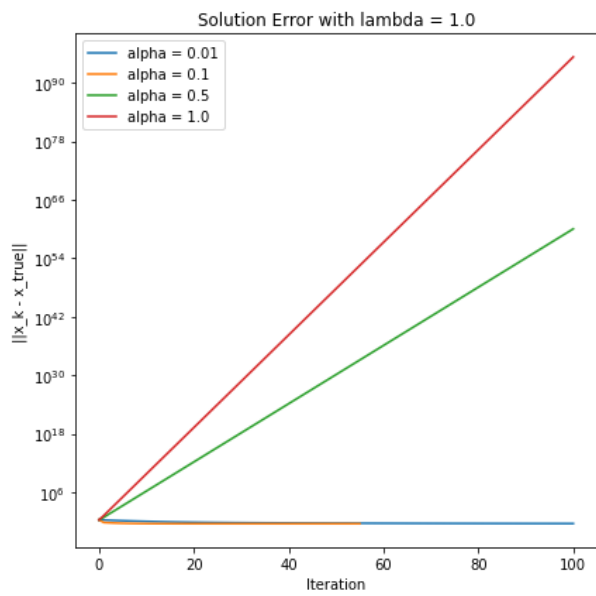
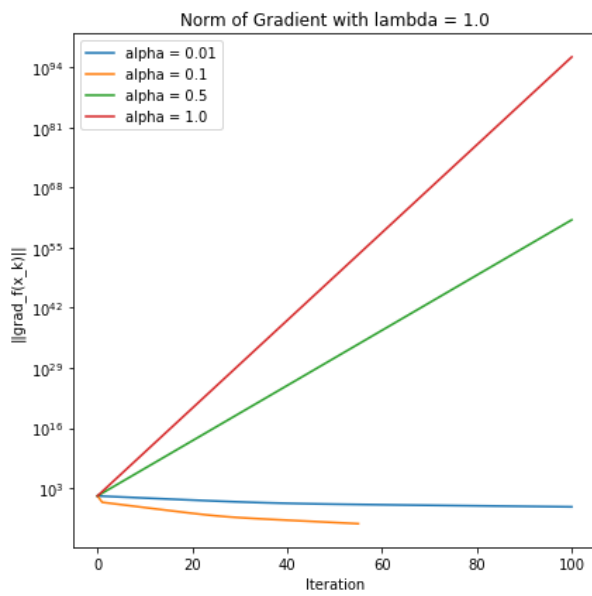
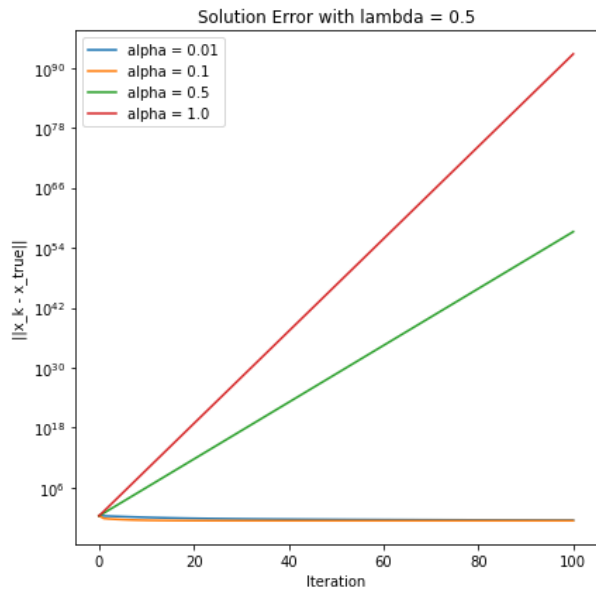
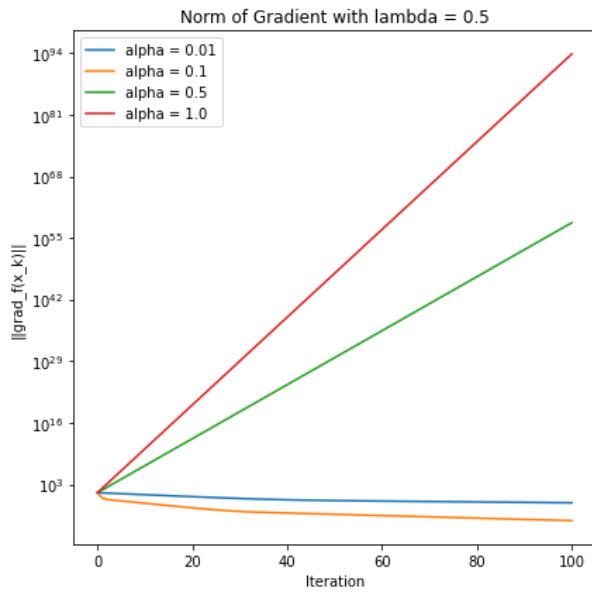


backtracking f4

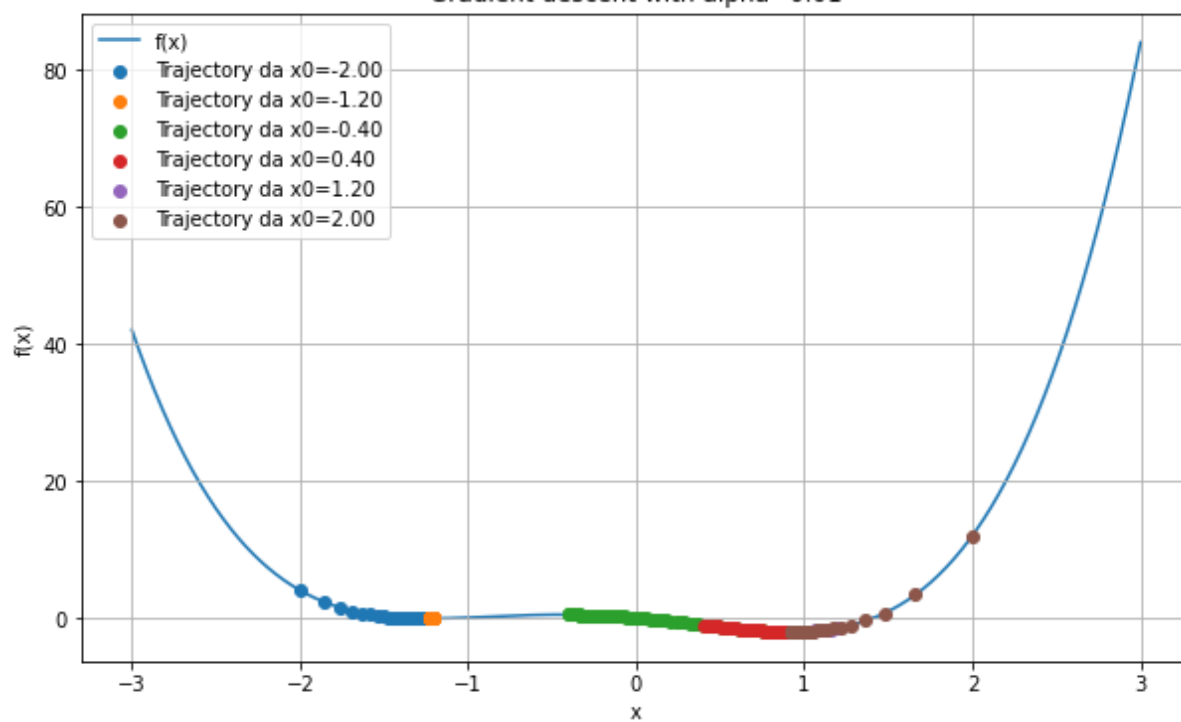
no backtracking f4



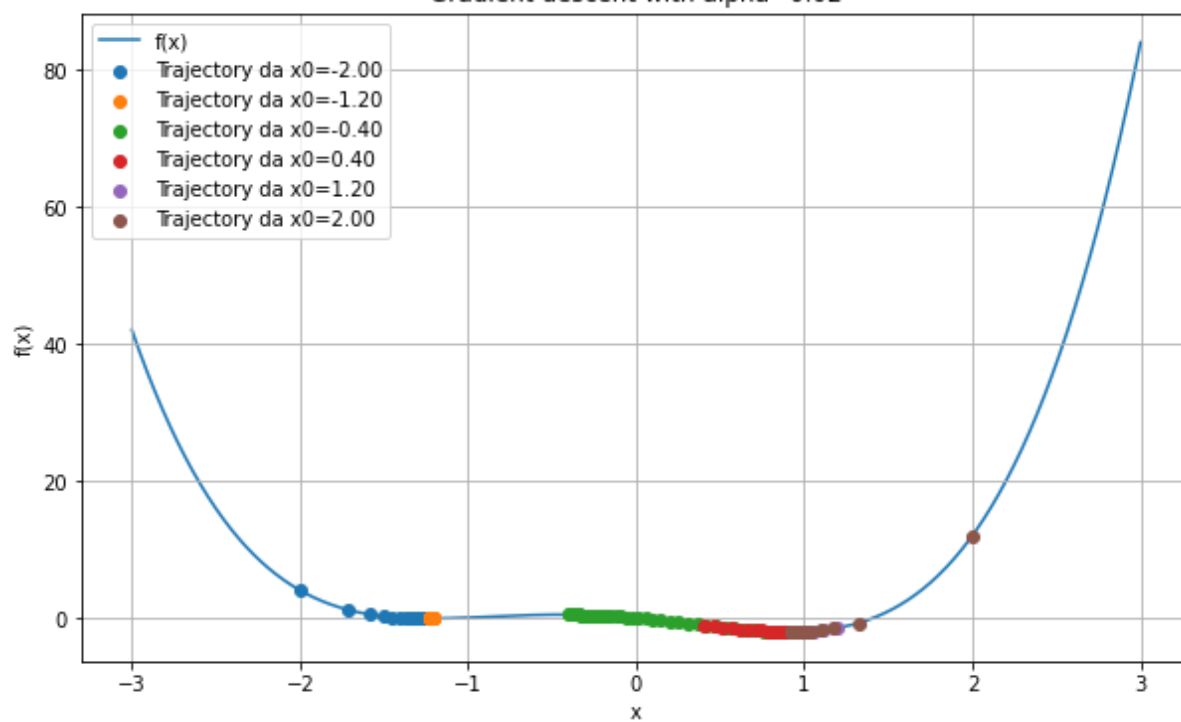




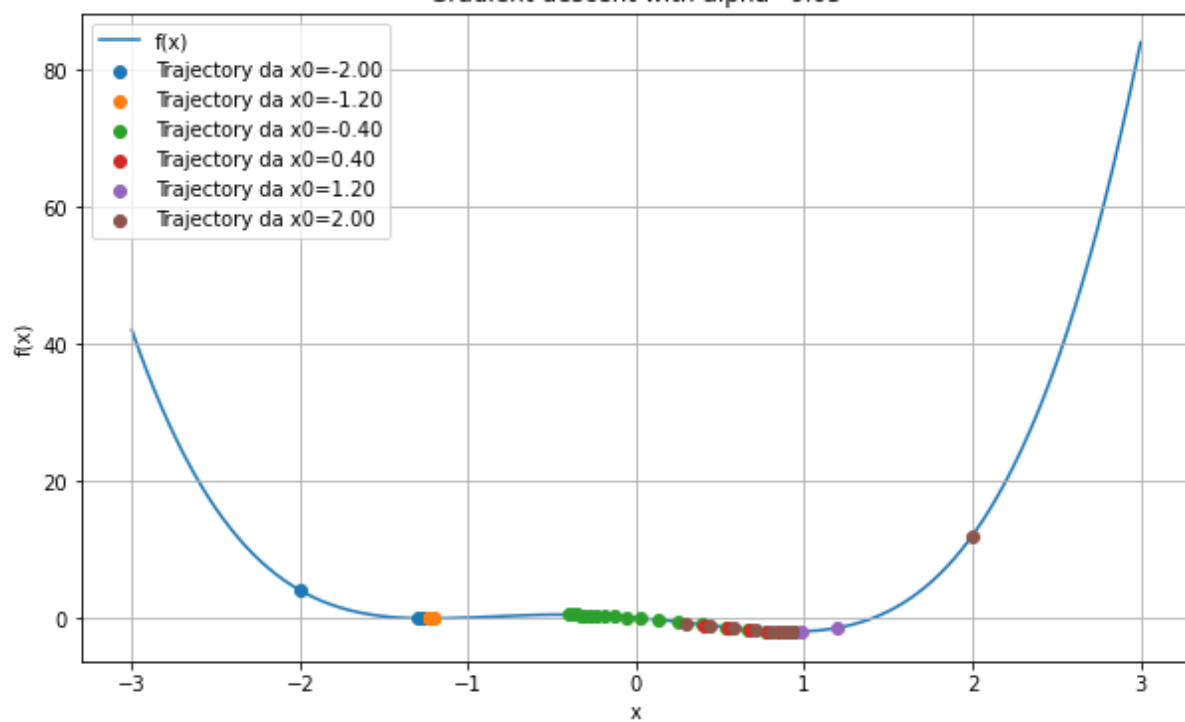
Gradient descent with $\alpha=0.01$

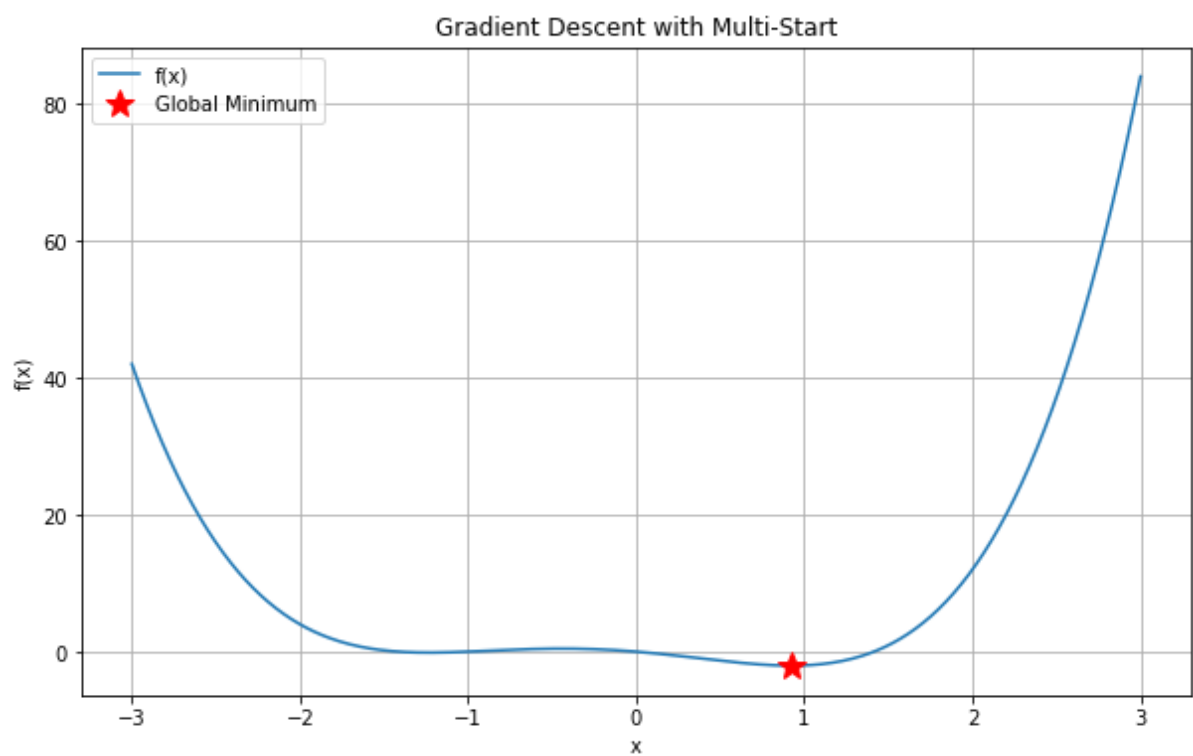
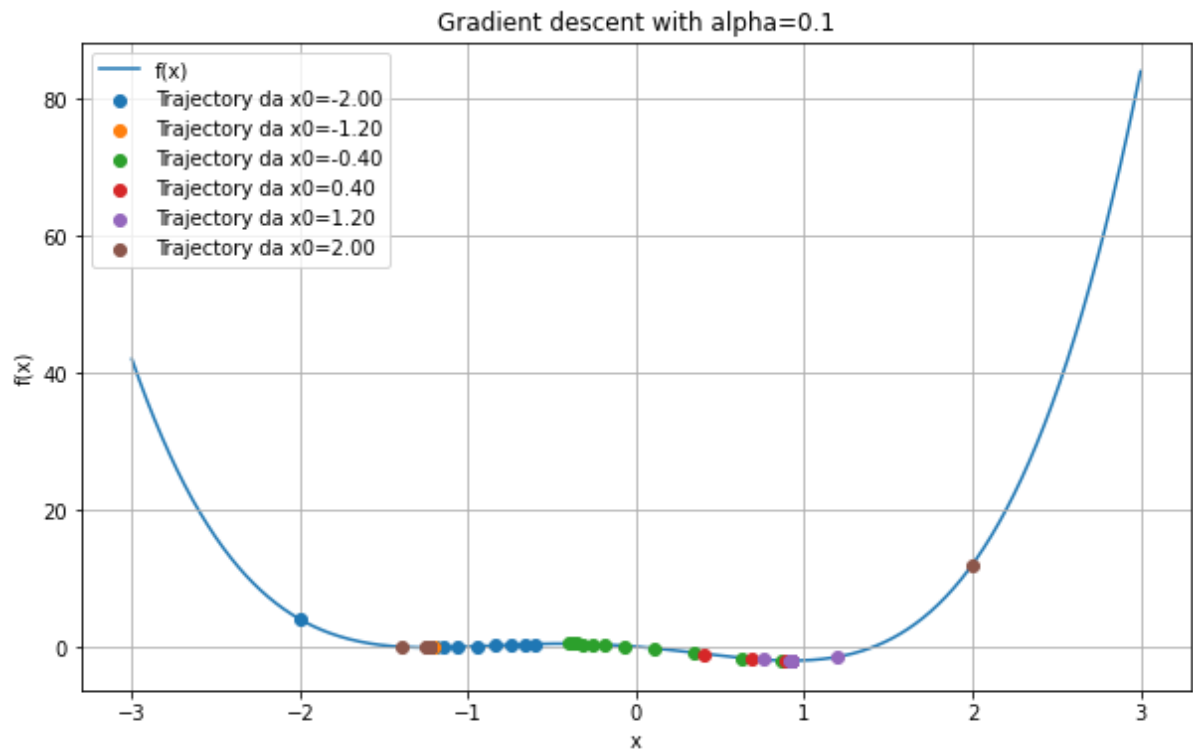


Gradient descent with $\alpha=0.02$



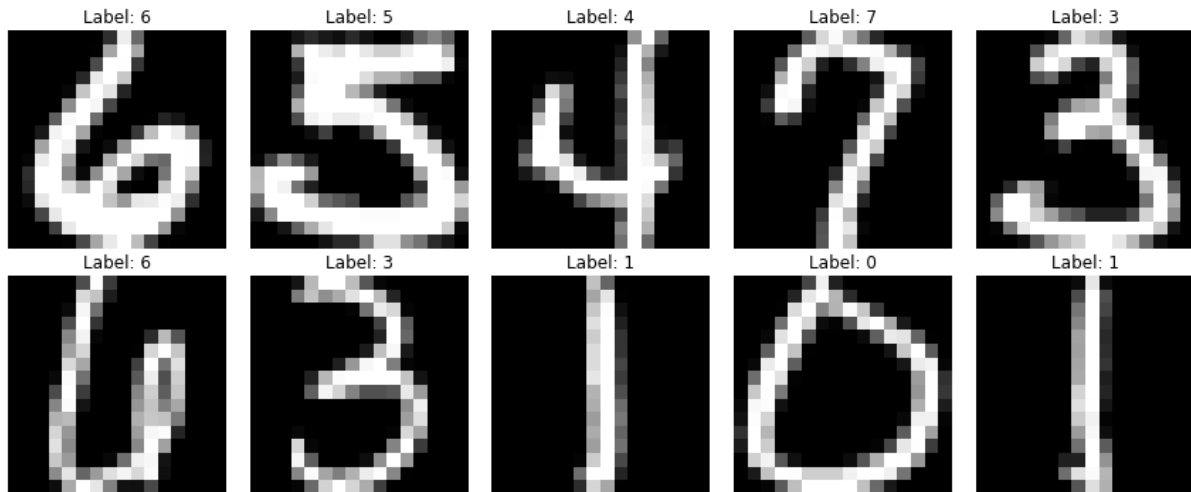
Gradient descent with alpha=0.05



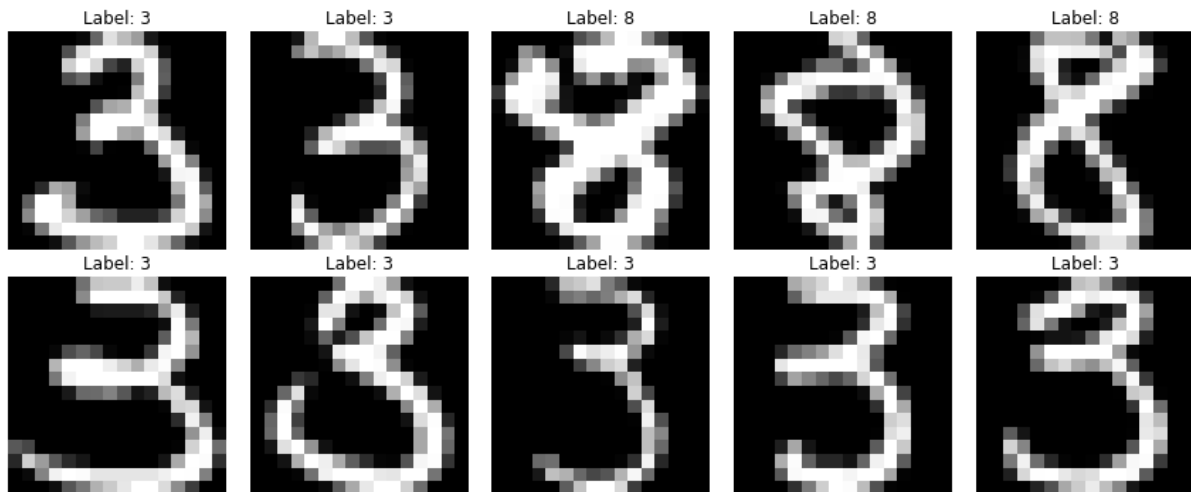


HomeWork 3_2

Shape of X (images): (256, 1707)
 Shape of I (labels): (1, 1707)



(256, 275) (1, 275)



(257, 220) (1, 220) (257, 55) (1, 55)

Accuracy: 1.0

Shape of X (images): (256, 1707)

Shape of I (labels): (1, 1707)

Filtered Data for digits (0, 1): X_filtered.shape = (257, 571), Y_filtered.shape = (1, 571)

Data Split: X_train.shape = (257, 285), Y_train.shape = (1, 285), X_test.shape = (257, 286),

Y_test.shape = (1, 286)

Binary Labels: [[0 1 1 1 0 1 0 1 0 0 0 1 0 0 0 1 0 1 0 0 1 1 1 1 0 0 0 1 1 1 0 1 0 1 1 0
1 0 0 0 0 0 0 0 0 0 1 1 0 1 0 0 1 0 0 1 0 0 1 0 1 0 1 0 0 0 0 0 0 1 0 0
0 1 1 1 0 0 1 1 1 0 1 0 0 0 1 0 0 1 0 1 1 0 1 1 1 0 0 0 0 1 0 0 0 1 0 0
1 1 0 1 1 0 0 0 1 1 0 1 1 0 1 0 0 0 1 1 1 0 0 1 1 1 0 1 0 0 0 0 1 1 1 0
0 0 0 0 1 0 1 0 1 0 0 1 1 1 1 1 0 1 0 1 1 0 1 0 1 1 0 1 1 0 0 0 0 1 0 0
1 1 0 0 1 1 1 0 0 1 0 1 1 0 0 1 0 0 0 0 0 1 1 1 1 0 0 0 0 1 0 1 1 0 0 1
0 1 1 1 0 1 1 1 0 1 0 0 0 1 0 0 1 0 1 1 0 0 0 1 0 1 1 0 0 1 0 0 0 0 0 0
0 0 1 0 0 0 1 0 0 0 1 1 1 1 0 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 0 0 0 1 1]]

Accuracy for digits 0, 1 with batch size 32 and training percentage 0.5: 0.9965034965034965

Binary Labels: [[0 1 1 1 0 1 0 1 0 0 0 1 0 0 0 1 0 1 0 0 1 1 1 1 0 0 0 1 1 1 0 1 0 1 1 0
1 0 0 0 0 0 0 0 0 0 1 1 0 1 0 0 1 0 0 1 0 0 1 0 1 0 1 0 0 0 0 0 0 1 0 0
0 1 1 1 0 0 1 1 1 0 1 0 0 0 1 0 0 1 0 1 1 0 1 1 1 0 0 0 0 1 0 0 0 1 0 0
1 1 0 1 1 0 0 0 1 1 0 1 1 0 1 0 0 0 1 1 1 0 0 1 1 1 0 1 0 0 0 0 1 1 1 0

1000000000110100100100101010000000100

0 1 1 1 0 0 1 1 1 0 1 0 0 0 1 0 0 1 0 1 1 0 1 1 1 0 0 0 0 1 0 0 0 1 0 0
1 1 0 1 1 0 0 0 1 1 0 1 1 0 1 0 0 0 1 1 1 0 0 1 1 1 0 1 0 0 0 0 1 1 1 0
0 0 0 0 1 0 1 0 1 0 0 1 1 1 1 1 0 1 0 1 1 0 1 0 1 1 0 1 1 0 0 0 0 1 0 0
1 1 0 0 1 1 1 0 0 1 0 1 1 0 0 1 0 0 0 0 0 1 1 1 1 0 0 0 0 1 0 1 1 0 0 1
0 1 1 1 0 1 1 1 0 1 0 0 0 1 0 0 1 0 1 1 0 0 0 1 0 1 1 0 0 1 0 0 0 0 0 0
0 0 1 0 0 0 1 0 0 0 1 1 1 1 0 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 0 0 1 1]]

Accuracy for digits 0, 1 with batch size 128 and training percentage 0.7: 0.9965034965034965

Binary Labels: [[0 1 1 1 0 1 0 1 0 1 0 0 0 1 0 0 0 1 0 1 0 0 1 1 1 1 0 0 0 1 1 1 0 1 0 1 1 0
1 0 0 0 0 0 0 0 0 0 1 1 0 1 0 0 1 0 0 1 0 0 1 0 1 0 1 0 0 0 0 0 0 1 0 0
0 1 1 1 0 0 1 1 1 0 1 0 0 0 1 0 0 1 0 1 1 0 1 1 1 0 0 0 0 1 0 0 0 1 0 0
1 1 0 1 1 0 0 0 1 1 0 1 1 0 1 0 0 0 1 1 1 0 0 1 1 1 0 1 0 0 0 0 1 1 1 0
0 0 0 0 1 0 1 0 1 0 0 1 1 1 1 1 0 1 0 1 1 0 1 0 1 1 0 1 1 0 0 0 0 1 0 0
1 1 0 0 1 1 1 0 0 1 0 1 1 0 0 1 0 0 0 0 0 1 1 1 1 0 0 0 0 1 0 1 1 0 0 1
0 1 1 1 0 1 1 1 0 1 0 0 0 1 0 0 1 0 1 1 0 0 0 1 0 1 1 0 0 1 0 0 0 0 0 0
0 0 1 0 0 0 1 0 0 0 1 1 1 1 0 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 0 0 1 1]]

Accuracy for digits 0, 1 with batch size 128 and training percentage 0.9: 0.9965034965034965

Data Split: X_train.shape = (257, 399), Y_train.shape = (1, 399), X_test.shape = (257, 172),
Y_test.shape = (1, 172)

Binary Labels: [[1 0 0 0 0 1 0 0 1 0 1 0 1 0 1 0 0 1 0 1 0 1 1 0 1 1 0 1 1 0 1 0 1 1 1 0
0 1 0 0 1 1 1 0 1 1 1 0 1 0 1 1 0 0 0 1 0 0 0 1 1 1 0 0 1 1 1 0 0 0 0 0
0 0 0 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 1 1 1 0 0 1 1 1 1 1 0 1 1 0 0 0 1
1 0 1 0 0 0 0 1 1 0 1 1 0 1 0 0 1 0 1 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0
0 1 0 0 0 1 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 0 1 1 0 0]]

Accuracy for digits 0, 1 with batch size 32 and training percentage 0.5: 0.9941860465116279

Binary Labels: [[1 0 0 0 0 1 0 0 1 0 1 0 1 0 1 0 0 1 0 1 0 1 1 0 1 1 0 1 1 0 1 0 1 1 1 0
0 1 0 0 1 1 1 0 1 1 1 0 1 0 1 1 0 0 0 1 0 0 0 1 1 1 0 0 1 1 1 0 0 0 0 0
0 0 0 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 1 1 1 0 0 1 1 1 1 1 0 1 1 0 0 0 1
1 0 1 0 0 0 0 1 1 0 1 1 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0
0 1 0 0 0 1 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 0 1 1 0 0]]

Accuracy for digits 0, 1 with batch size 32 and training percentage 0.7: 0.9941860465116279

Binary Labels: [[1 0 0 0 0 1 0 0 1 0 1 0 1 0 1 0 0 1 0 1 0 1 1 0 1 1 0 1 1 0 1 0 1 1 1 0
0 1 0 0 1 1 1 0 1 1 1 0 1 0 1 1 0 0 0 1 0 0 0 1 1 1 0 0 1 1 1 0 0 0 0 0
0 0 0 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 1 1 1 0 0 1 1 1 1 1 0 1 1 0 0 0 1
1 0 1 0 0 0 0 1 1 0 1 1 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0
0 1 0 0 0 1 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 0 1 1 0 0]]

Accuracy for digits 0, 1 with batch size 32 and training percentage 0.9: 0.9941860465116279

Binary Labels: [[1 0 0 0 0 1 0 0 1 0 1 0 1 0 1 0 0 1 0 1 0 1 1 0 1 1 0 1 1 0 1 0 1 1 1 0
0 1 0 0 1 1 1 0 1 1 1 0 1 0 1 1 0 0 0 1 0 0 0 1 1 1 0 0 1 1 1 0 0 0 0 0
0 0 0 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 1 1 1 0 0 1 1 1 1 1 0 1 1 0 0 0 1
1 0 1 0 0 0 0 1 1 0 1 1 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0
0 1 0 0 0 1 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 0 1 1 0 0]]

Accuracy for digits 0, 1 with batch size 64 and training percentage 0.5: 0.9941860465116279

Binary Labels: [[1 0 0 0 0 1 0 0 1 0 1 0 1 0 1 0 0 1 0 1 0 1 1 0 1 1 0 1 1 0 1 0 1 1 1 0
0 1 0 0 1 1 1 0 1 1 1 0 1 0 1 1 0 0 0 1 0 0 0 1 1 1 0 0 1 1 1 0 0 0 0 0
0 0 0 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 1 1 1 0 0 1 1 1 1 1 0 1 1 0 0 0 1
1 0 1 0 0 0 0 1 1 0 1 1 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0
0 1 0 0 0 1 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 0 1 1 0 0]]

Accuracy for digits 0, 1 with batch size 64 and training percentage 0.7: 0.9941860465116279

Binary Labels: [[1 0 0 0 0 1 0 0 1 0 1 0 1 0 1 0 0 1 0 1 0 1 1 0 1 1 0 1 1 0 1 0 1 1 1 0
0 1 0 0 1 1 1 0 1 1 1 0 1 0 1 1 0 0 0 1 0 0 0 1 1 1 0 0 1 1 1 0 0 0 0 0
0 0 0 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 1 1 1 0 0 1 1 1 1 1 0 1 1 0 0 0 1
1 0 1 0 0 0 0 1 1 0 1 1 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0
1 0 1 0 0 0 0 1 1 0 1 1 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0

0 1 0 0 0 1 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 0 1 1 0 0]]

Accuracy for digits 0, 1 with batch size 64 and training percentage 0.9: 0.9941860465116279

Binary Labels: [[1 0 0 0 0 1 0 0 1 0 1 0 1 0 1 0 0 1 0 1 0 1 1 0 1 1 0 1 0 1 1 1 0

0 1 0 0 1 1 1 0 1 1 1 0 1 0 1 1 0 0 0 1 0 0 0 1 1 1 0 0 1 1 1 0 0 0 0 0

0 0 0 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 1 1 1 0 0 1 1 1 1 1 0 1 1 0 0 0 1

1 0 1 0 0 0 0 1 1 0 1 1 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0

0 1 0 0 0 1 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 0 1 1 0 0]]

Accuracy for digits 0, 1 with batch size 128 and training percentage 0.5: 0.9941860465116279

Binary Labels: [[1 0 0 0 0 1 0 0 1 0 1 0 1 0 1 0 0 1 0 1 0 1 1 0 1 1 0 1 1 0 1 0 1 1 1 0

0 1 0 0 1 1 1 0 1 1 1 0 1 0 1 1 0 0 0 1 0 0 0 1 1 1 0 0 1 1 1 0 0 0 0 0

0 0 0 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 1 1 1 0 0 1 1 1 1 1 0 1 1 0 0 0 1

1 0 1 0 0 0 0 1 1 0 1 1 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0

0 1 0 0 0 1 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 0 1 1 0 0]]

Accuracy for digits 0, 1 with batch size 128 and training percentage 0.7: 0.9941860465116279

Binary Labels: [[1 0 0 0 0 1 0 0 1 0 1 0 1 0 1 0 0 1 0 1 0 1 1 0 1 1 0 1 1 0 1 0 1 1 1 0

0 1 0 0 1 1 1 0 1 1 1 0 1 0 1 1 0 0 0 1 0 0 0 1 1 1 0 0 1 1 1 0 0 0 0 0

0 0 0 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 1 1 1 0 0 1 1 1 1 1 0 1 1 0 0 0 1

1 0 1 0 0 0 0 1 1 0 1 1 0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0

0 1 0 0 0 1 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 0 1 1 0 0]]

Accuracy for digits 0, 1 with batch size 128 and training percentage 0.9: 0.9941860465116279

Data Split: X_train.shape = (257, 513), Y_train.shape = (1, 513), X_test.shape = (257, 58),

Y_test.shape = (1, 58)

Binary Labels: [[0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0 0 0 0 1 0 0 1 1 0 0 1 0 1 1 1 0 0 0 1

0 1 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 1 0 0 1]]

Accuracy for digits 0, 1 with batch size 32 and training percentage 0.5: 1.0

Binary Labels: [[0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0 0 0 0 1 0 0 1 1 0 0 1 0 1 1 1 0 0 0 1

0 1 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 1 0 0 1]]

Accuracy for digits 0, 1 with batch size 32 and training percentage 0.7: 1.0

Binary Labels: [[0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0 0 0 0 1 0 0 1 1 0 0 1 0 1 1 1 0 0 0 1

0 1 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 1 0 0 1]]

Accuracy for digits 0, 1 with batch size 32 and training percentage 0.9: 1.0

Binary Labels: [[0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0 0 0 0 1 0 0 1 1 0 0 1 0 1 1 1 0 0 0 1

0 1 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 1 0 0 1]]

Accuracy for digits 0, 1 with batch size 64 and training percentage 0.5: 1.0

Binary Labels: [[0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0 0 0 0 1 0 0 1 1 0 0 1 0 1 1 1 0 0 0 1

0 1 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 1 0 0 1]]

Accuracy for digits 0, 1 with batch size 64 and training percentage 0.7: 1.0

Binary Labels: [[0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0 0 0 0 1 0 0 1 1 0 0 1 0 1 1 1 0 0 0 1

0 1 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 1 0 0 1]]

Accuracy for digits 0, 1 with batch size 64 and training percentage 0.9: 1.0

Binary Labels: [[0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0 0 0 0 1 0 0 1 1 0 0 1 0 1 1 1 0 0 0 1

0 1 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 1 0 0 1]]

Accuracy for digits 0, 1 with batch size 128 and training percentage 0.5: 1.0

Binary Labels: [[0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0 0 0 0 1 0 0 1 1 0 0 1 0 1 1 1 0 0 0 1

0 1 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 1 0 0 1]]

Accuracy for digits 0, 1 with batch size 128 and training percentage 0.7: 1.0

Binary Labels: [[0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 1 0 0 0 0 1 0 0 1 1 0 0 1 0 1 1 1 0 0 0 1

0 1 0 0 1 0 1 0 1 0 0 0 1 0 0 0 0 0 1 0 0 1]]

Accuracy for digits 0, 1 with batch size 128 and training percentage 0.9: 1.0

Filtered Data for digits (2, 3): X_filtered.shape = (257, 333), Y_filtered.shape = (1, 333)

Data Split: X_train.shape = (257, 166), Y_train.shape = (1, 166), X_test.shape = (257, 167),

Y_test.shape = (1, 167)

Binary Labels: [[1 0 0 0 1 1 0 1 0 1 0 0 1 1 0 0 0 0 0 1 0 0 0 1 0 0 0 0 0 0 0 0 1 1 1 0
0 0 1 1 0 0 1 0 0 0 0 0 1 0 1 0 0 1 0 0 1 1 0 0 0 0 0 0 0 1 0 0 0 1 1
1 1 1 0 1 1 1 0 1 0 0 1 0 0 1 0 0 0 0 0 1 1 1 0 1 0 1 0 1 1 1 1 1 0 1 0
0 1 1 0 1 0 0 0 0 0 1 0 0 0 0 1 0 1 1 0 1 1 1 0 1 1 0 1 0 0 0 0 1 0 0 1

Binary Labels: [[0 0 1 1 1 1 1 1 1 0 0 0 1 0 0 0 0 0 1 1 1 0 0 0 0 0 0 0 1 0 1 0 0 1]]
Accuracy for digits 2, 3 with batch size 64 and training percentage 0.9: 0.4117647058823529
Binary Labels: [[0 0 1 1 1 1 1 1 1 1 0 0 0 1 0 0 0 0 0 1 1 1 0 0 0 0 0 0 0 1 0 1 0 0 1]]
Accuracy for digits 2, 3 with batch size 128 and training percentage 0.5: 0.4117647058823529
Binary Labels: [[0 0 1 1 1 1 1 1 1 1 0 0 0 1 0 0 0 0 0 1 1 1 0 0 0 0 0 0 0 1 0 1 0 0 1]]
Accuracy for digits 2, 3 with batch size 128 and training percentage 0.7: 0.4117647058823529
Binary Labels: [[0 0 1 1 1 1 1 1 1 1 0 0 0 1 0 0 0 0 0 1 1 1 0 0 0 0 0 0 0 1 0 1 0 0 1]]
Accuracy for digits 2, 3 with batch size 128 and training percentage 0.9: 0.4117647058823529
Filtered Data for digits (4, 5): X_filtered.shape = (257, 210), Y_filtered.shape = (1, 210)
Data Split: X_train.shape = (257, 105), Y_train.shape = (1, 105), X_test.shape = (257, 105),
Y_test.shape = (1, 105)
Binary Labels: [[0 0 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 1 0 0 0 0 0 1 1 1 0 1 0 1
0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 1 0 0 0 0 0 1 1 1 0 1 0 1 1 1 0 1 0 0 1 1
1 0 1 0 0 0 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 1 0]]
Accuracy for digits 4, 5 with batch size 32 and training percentage 0.5: 0.3904761904761905
Binary Labels: [[0 0 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 1 0 0 0 0 0 1 1 1 0 1 0 1
0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 1 0 0 0 0 0 1 1 1 0 1 0 1 1 1 0 1 0 0 1 1
1 0 1 0 0 0 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 1 0]]
Accuracy for digits 4, 5 with batch size 32 and training percentage 0.7: 0.3904761904761905
Binary Labels: [[0 0 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 1 0 0 0 0 0 1 1 1 0 1 0 1
0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 1 0 0 0 0 0 1 1 1 0 1 0 1 1 1 0 1 0 0 1 1
1 0 1 0 0 0 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 1 0]]
Accuracy for digits 4, 5 with batch size 32 and training percentage 0.9: 0.3904761904761905
Binary Labels: [[0 0 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 1 0 0 0 0 0 1 1 1 0 1 0 1
0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 1 0 0 0 0 0 1 1 1 0 1 0 1 1 1 0 1 0 0 1 1
1 0 1 0 0 0 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 1 0]]
Accuracy for digits 4, 5 with batch size 64 and training percentage 0.5: 0.3904761904761905
Binary Labels: [[0 0 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 1 0 0 0 0 0 1 1 1 0 1 0 1
0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 1 0 0 0 0 0 1 1 1 0 1 0 1 1 1 0 1 0 0 1 1
1 0 1 0 0 0 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 1 0]]
Accuracy for digits 4, 5 with batch size 64 and training percentage 0.7: 0.3904761904761905
Binary Labels: [[0 0 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 1 0 0 0 0 0 1 1 1 0 1 0 1
0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 1 0 0 0 0 0 1 1 1 0 1 0 1 1 1 0 1 0 0 1 1
1 0 1 0 0 0 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 1 0]]
Accuracy for digits 4, 5 with batch size 64 and training percentage 0.9: 0.3904761904761905
Binary Labels: [[0 0 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 1 0 0 0 0 0 1 1 1 0 1 0 1
0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 1 0 0 0 0 0 1 1 1 0 1 0 1 1 1 0 1 0 0 1 1
1 0 1 0 0 0 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 1 0]]
Accuracy for digits 4, 5 with batch size 128 and training percentage 0.5: 0.3904761904761905
Binary Labels: [[0 0 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 1 0 0 0 0 0 1 1 1 0 1 0 1
0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 1 0 0 0 0 0 1 1 1 0 1 0 1 1 1 0 1 0 0 1 1
1 0 1 0 0 0 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 1 0]]
Accuracy for digits 4, 5 with batch size 128 and training percentage 0.7: 0.3904761904761905
Binary Labels: [[0 0 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 1 0 1 0 0 0 0 0 1 1 1 0 1 0 1
0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 1 0 0 0 0 0 1 1 1 0 1 0 1 1 1 0 1 0 0 1 1
1 0 1 0 0 0 0 0 0 0 0 1 1 1 0 1 1 1 0 0 0 0 0 1 0 1 1 0 0 0 1 1 0]]
Accuracy for digits 4, 5 with batch size 128 and training percentage 0.9: 0.3904761904761905
Data Split: X_train.shape = (257, 147), Y_train.shape = (1, 147), X_test.shape = (257, 63),
Y_test.shape = (1, 63)
Binary Labels: [[1 0 0 1 1 0 0 0 1 1 0 0 0 0 0 0 1 0 1 1 0 0 1 1 0 1 0 0 0 0 0 0 0 1 1 1
1 1 1 0 0 1 1 1 1 0 1 1 0 1 1 1 0 0 0 1 0 1 0 1 1 0 0]]
Accuracy for digits 4, 5 with batch size 32 and training percentage 0.5: 0.47619047619047616

ACCURACY

{(0, 1): {32: {0.5: 1.0, 0.7: 1.0, 0.9: 1.0}, 64: {0.5: 1.0, 0.7: 1.0, 0.9: 1.0}, 128: {0.5: 1.0, 0.7: 1.0, 0.9: 1.0}},
(2, 3): {32: {0.5: 0.4117647058823529, 0.7: 0.4117647058823529, 0.9: 0.4117647058823529}, 64: {0.5: 0.4117647058823529, 0.7: 0.4117647058823529, 0.9: 0.4117647058823529}, 128: {0.5: 0.4117647058823529, 0.7: 0.4117647058823529, 0.9: 0.4117647058823529}},
(4, 5): {32: {0.5: 0.38095238095238093, 0.7: 0.38095238095238093, 0.9: 0.38095238095238093}, 64: {0.5: 0.38095238095238093, 0.7: 0.38095238095238093, 0.9: 0.38095238095238093}, 128: {0.5: 0.38095238095238093, 0.7: 0.38095238095238093, 0.9: 0.38095238095238093}}}}