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In [1]: 1 import numpy as np
2 import pandas as pd
3 import matplotlib.pyplot as plt
4 %matplotlib inline

In [2]: 1 def f(x):
2     return(100 * (x[1][0] - x[0][0]**2)**2 + (1 - x[0][0])**2)
3
4 def grad_f(x):
5     return(np.array([[400 * x[0][0]**3 - 400 * x[0][0] * x[1][0] + 2 * x[0][0] - 2], [200 * (x[1][0] - x[0][0]**2)]]))
6
7 def hessian_f(x):
8     return(np.array([[1200 * x[0][0]**2 - 400 * x[1][0] + 2, -400 * x[0][0]], [-400 * x[0][0], 200]]))

In [3]: 1 def back_track(x, alpha, c, d, rho):
2     while True:
3         if f(x + alpha * d) <= f(x) + c * alpha * grad_f(x).T @ d:
4             return alpha
5             alpha *= rho

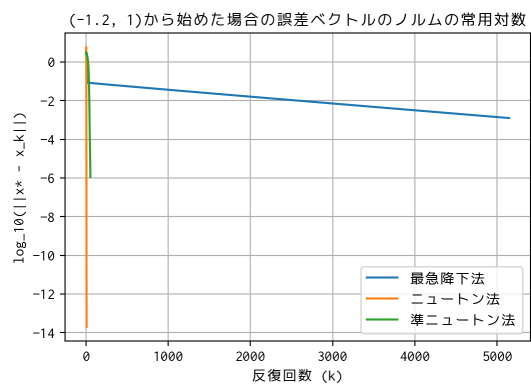
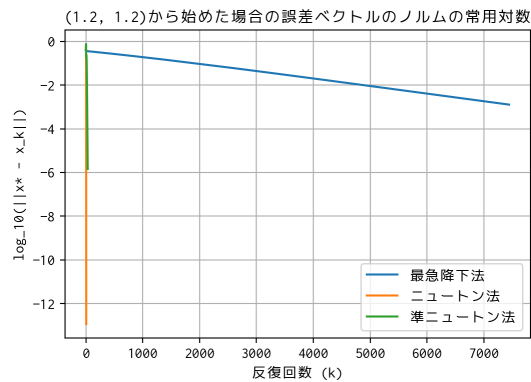
In [4]: 1 tol = 1e-6

In [5]: 1 def gradient_descent(x_0, alpha = 0.5, c = 0.1, rho = 0.8, n_itr = 100):
2     xs = np.array([x_0])
3     error = tol + 1
4     while error > tol:
5         d = -1 * grad_f(xs[-1])
6         xs = np.append(xs, [xs[-1] + back_track(xs[-1], alpha, c, d, rho) * d], axis = 0)
7         error = np.linalg.norm(xs[-1] - xs[-2]) / np.linalg.norm(xs[-1])
8     return xs
9
10 def newton_method(x_0, n_itr = 50):
11     xs = np.array([x_0])
12     error = tol + 1
13     while error > tol:
14         d = -1 * np.linalg.inv(hessian_f(xs[-1])) @ grad_f(xs[-1])
15         xs = np.append(xs, [xs[-1] + d], axis = 0)
16         error = np.linalg.norm(xs[-1] - xs[-2]) / np.linalg.norm(xs[-1])
17     return xs
18
19 def quasi_newton_method(x_0, H = np.array([[1, 0], [0, 1]]), alpha = 0.5, c = 0.1, rho = 0.8, n_itr = 50):
20     xs = np.array([x_0])
21     error = tol + 1
22     while error > tol:
23         d = -1 * H @ grad_f(xs[-1])
24         xs = np.append(xs, [xs[-1] + back_track(xs[-1], alpha, c, d, rho) * d], axis = 0)
25         s = xs[-1] - xs[-2]
26         y = grad_f(xs[-1]) - grad_f(xs[-2])
27         H = (np.identity(2) - (s @ y.T) / (y.T @ s)) @ H @ (np.identity(2) - (y @ s.T) / (y.T @ s)) + (s @ s.T) / (s.T @ y)
28         error = np.linalg.norm(xs[-1] - xs[-2]) / np.linalg.norm(xs[-1])
29     return xs

In [6]: 1 def error(v, optimum_point = np.array([[1], [1]])):
2     return np.log10(np.linalg.norm(v - optimum_point))
3
4 def plot_errors(start, start_str):
5     gradient_transition = gradient_descent(start)
6     newton_transition = newton_method(start)
7     quasi_newton_transition = quasi_newton_method(start)
8
9     plt.plot(np.arange(len(gradient_transition)), np.apply_along_axis(error, 1, gradient_transition).flatten(), label='最急降下法')
10    plt.plot(np.arange(len(newton_transition)), np.apply_along_axis(error, 1, newton_transition).flatten(), label='ニュートン法')
11    plt.plot(np.arange(len(quasi_newton_transition)), np.apply_along_axis(error, 1, quasi_newton_transition).flatten(), label='準ニュートン法')
12    plt.legend()
13    plt.title(start_str + "から始めた場合の誤差ベクトルのノルムの常用対数")
14    plt.xlabel('反復回数 (k)')
15    plt.ylabel('log_10(||x* - x_k||)')
16    plt.grid()
17    plt.show()

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In [7]: 1 plot_errors(np.array([[1.2], [1.2]]), '(1.2, 1.2)')
2 plot_errors(np.array([[-1.2], [1.]]), '(-1.2, 1)')
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In [8]: 1 def plot_transitions(start, start_str):
2     plt.scatter(start[0], start[1], label='開始点', color='blue')
3     plt.scatter([1], [1], label='最適点', color='red')
4
5     gradient_transition = gradient_descent(start)
6     newton_transition = newton_method(start)
7     quasi_newton_transition = quasi_newton_method(start)
8
9     plt.plot(gradient_transition[:, 0], gradient_transition[:, 1], label='最急降下法')
10    plt.plot(newton_transition[:, 0], newton_transition[:, 1], label='ニュートン法')
11    plt.plot(quasi_newton_transition[:, 0], quasi_newton_transition[:, 1], label='準ニュートン法')
12    plt.legend()
13    plt.title(start_str + "から始めた場合のベクトルの推移")
14    plt.xlabel('x_1')
15    plt.ylabel('x_2')
16    plt.grid()
17    plt.show()
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In [9]: 1 plot_transitions(np.array([[1.2], [1.2]]), '(1.2, 1.2)')
2 plot_transitions(np.array([[-1.2], [1.]]), '(-1.2, 1)')
```

