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In [1]: 1 import numpy as np
2 import pandas as pd
3 import matplotlib.pyplot as plt
4 import matplotlib
5 font = {"family": "AppleGothic"}
6 matplotlib.rc("font", **font)
7 % matplotlib inline

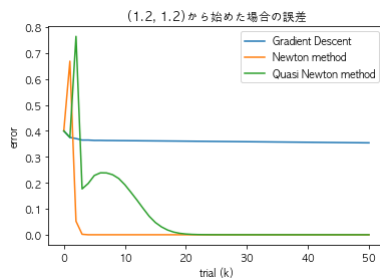
In [2]: 1 def f(x):
2     return(100 * (x[1][0] - x[0][0] ** 2) ** 2 + (1 - x[0][0]) ** 2)
3
4 def grad_f(x):
5     return(np.array([[400 * x[0][0] ** 3 - 400 * x[0][0] * x[1][0] + 2 * x[0][0] - 2], [200 * (x[1][0] - x[0][0] ** 2)]]))
6
7 def hessian_f(x):
8     return(np.array([[1200 * x[0][0] ** 2 - 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 def gradient_descent(x_0, alpha, c, rho, n_itr):
2     xs = np.array([x_0])
3     for k in range(n_itr):
4         d = -1 * grad_f(xs[k])
5         xs = np.append(xs, [xs[k] + back_track(xs[k], alpha, c, d, rho) * d], axis = 0)
6     return xs
7
8 def newton_method(x_0, n_itr):
9     xs = np.array([x_0])
10    for k in range(n_itr):
11        d = -1 * np.linalg.inv(hessian_f(xs[k])) @ grad_f(xs[k])
12        xs = np.append(xs, [xs[k] + d], axis = 0)
13    return xs
14
15 def quasi_newton_method(x_0, H, alpha, c, rho, n_itr):
16    xs = np.array([x_0])
17    for k in range(n_itr):
18        d = -1 * H @ grad_f(xs[k])
19        xs = np.append(xs, [xs[k] + back_track(xs[k], alpha, c, d, rho) * d], axis = 0)
20        s = xs[k + 1] - xs[k]
21        y = grad_f(xs[k + 1]) - grad_f(xs[k])
22        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)
23    return xs

In [5]: 1 gradient_transition = gradient_descent(np.array([[1.2], [1.2]]), 0.5, 0.1, 0.8, 50)
2 newton_transition = newton_method(np.array([[1.2], [1.2]]), 50)
3 quasi_newton_transition = quasi_newton_method(np.array([[1.2], [1.2]]), np.array([[1, 0], [0, 1]]), 0.5, 0.1, 0.8, 50)
4
5 optimum_point = np.array([[1], [1]])
6 def error(v):
7     return np.linalg.norm(v - optimum_point)
8
9 cnt = np.arange(51)
10 plt.plot(cnt, np.apply_along_axis(error, 1, gradient_transition).flatten(), label='Gradient Descent')
11 plt.plot(cnt, np.apply_along_axis(error, 1, newton_transition).flatten(), label='Newton method')
12 plt.plot(cnt, np.apply_along_axis(error, 1, quasi_newton_transition).flatten(), label='Quasi Newton method')
13 plt.legend()
14 plt.title('(1.2, 1.2)から始めた場合の誤差')
15 plt.xlabel('trial (k)')
16 plt.ylabel('error')
17 plt.show()

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In [6]: 1 gradient_transition = gradient_descent(np.array([[-1.2], [1.]]), 0.5, 0.1, 0.8, 50)
2 newton_transition = newton_method(np.array([[-1.2], [1.]]), 50)
3 quasi_newton_transition = quasi_newton_method(np.array([[-1.2], [1.]]), np.array([[1, 0], [0, 1]]), 0.5, 0.1, 0.8, 50)
4
5 optimum_point = np.array([[1], [1]])
6 def error(v):
7     return np.linalg.norm(v - optimum_point)
8
9 cnt = np.arange(51)
10 plt.plot(cnt, np.apply_along_axis(error, 1, gradient_transition).flatten(), label='Gradient Descent')
11 plt.plot(cnt, np.apply_along_axis(error, 1, newton_transition).flatten(), label='Newton method')
12 plt.plot(cnt, np.apply_along_axis(error, 1, quasi_newton_transition).flatten(), label='Quasi Newton method')
13 plt.legend()
14 plt.title('(-1.2, 1)から始めた場合の誤差')
15 plt.xlabel('trial (k)')
16 plt.ylabel('error')
17 plt.show()

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