# EECE 5550 HW3 Problem 1: Optimization Methods

To begin, make a copy of this notebook so you can save changes.

After completing these problems, please save this notebook as a pdf and combine it with your solutions to the written problems so that you can upload a single pdf to Gradescope.

▼ These first few cells just implement some methods you may find useful

```
import numpy as np
import matplotlib.pyplot as plt
from typing import Callable
# Draw 2 plots:
   # On the left subplot, show:
   # - the function f(x) as contours
  # - the values of x as we search for the minimizer
   # On the right subplot, show:
   # - value of f(x) vs. current xi at each iteration of the optimization
   plt.subplots(1, 2)
   plt.subplot(1, 2, 2)
    z = f(xs)
   {\tt plt.\,plot(np.\,arange(len(z)),\quad z)}
   plt.xlabel('Iteration')
   plt.ylabel('f(x)')
   plt.gca().set_yscale('log')
   plt.subplot(1, 2, 1)
   # Optimization
   \texttt{plt.plot}(\texttt{xs}[:, \quad 0], \quad \texttt{xs}[:, \quad 1], \quad '-\texttt{x'})
   x_high = 2
   y_low = -2
   y_high = 2
   # Contour Plot
   num_xs = 20
   num_ys = 20
   x_vals = np.linspace(x_low, x_high, num_xs)
   y_vals = np.linspace(y_low, y_high, num_ys)
   X, Y = np.meshgrid(x_vals, y_vals)
   xy_pairs = np.vstack([X.ravel(), Y.ravel()]).T
   z = f(xy_pairs, k=k)
   Z = z.reshape(num_xs, num_ys)
   plt.gca().contour(X, Y, Z)
   plt.xlim([x_low, x_high])
   plt.ylim([y_low, y_high])
   plt. show()
def f(x: np.ndarray, k: int = 1) \rightarrow float:
  # f(x) = x^2 - xy + ky^2
  if x.ndim == 1:
      return x[0]**2 - x[0]*x[1] + k*x[1]**2
   elif x.ndim = 2:
      return x[:, 0]**2 - x[:, 0]*x[:, 1] + k*x[:, 1]**2
```

#### → 1.1 [2pts] Implement the gradient and hessian of f(x; k)

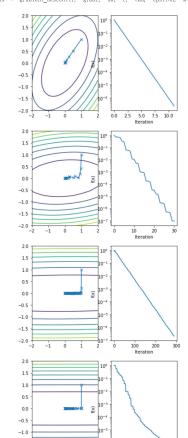
```
def gradf(x: np.ndarray, k: int = 1) → np.ndarray:
    return np.array([2 * x[0] - x[1], -x[0] + 2 * k * x[1]])
    def hessf(x: np.ndarray, k: int = 1) → np.ndarray:
    return np.array([2, -1], [-1, 2 * k]])
```

#### ▼ 1.2 [2pts] Implement Gradient Descent

```
x = xs[-1]
p = _gradf(x, k)
if np.linalg.norm(p) < epsilon:
    break
    alpha = 1
    while f(x + alpha*p, k) > f(x, k) - c*alpha*ap.linalg.norm(p)**2:
        alpha *= tau
        xs.append(x + alpha*p)
    xs = np.array(xs)
if plot:
    draw_plot(xs, f, k=k)
return xs
```

#### $ilde{\ \ }$ 1.2 [1pt] Run your gradient descent for various values of $\kappa$

```
 \begin{aligned} x0 &= \text{np.array([i., \ 1.])} \\ c &= 0.5 \\ \text{tau} &= 0.5 \\ \text{epsilon} &= 0.001 \\ \text{xs} &= \text{gradient_descent(f, gradf, x0, c, tau, epsilon, k=1, plot=True)} \\ \text{xs} &= \text{gradient_descent(f, gradf, x0, c, tau, epsilon, k=10, plot=True)} \\ \text{xs} &= \text{gradient_descent(f, gradf, x0, c, tau, epsilon, k=100, plot=True)} \\ \text{xs} &= \text{gradient_descent(f, gradf, x0, c, tau, epsilon, k=100, plot=True)} \\ \end{aligned}
```



For full credit on this problem, add some text description of what you observe about these results and why that is occurring (replace this text cell with your own thoughts).

## ▼ 1.4 [2pts] Implement Newton's Method

50 100 Iteration

```
def newton(
f: Callable,
gradf: Callable,
hessf: Callable,
x0: np.ndarray,
epsilon: float,
```

-1.5

```
plot: bool = False,
    k: int = 1
) -> np.ndarray:
    xs = [x0]
    while True:
        i x = xs[-1]
        grad = gradf(x, k=k)
        hess = hessf(x, k=k)
        if np.linalg.norm(grad) < epsilon:
            break
        xs.append(x - np.linalg.inv(hess) @ grad)
        is = np.array(xs)
        if plot:
            draw_plot(xs, f, k=k)
        return xs
```

### ullet 1.5 [1pt] Run your Newton's method for various values of $\kappa$

```
x0 = np.array([1, 1.])
epsilon = 0.001
xs = newton(f, gradf, hessf, x0, epsilon, k=1, plot=True)
xs = newton(f, gradf, hessf, x0, epsilon, k=10, plot=True)
xs = newton(f, gradf, hessf, x0, epsilon, k=100, plot=True)
xs = newton(f, gradf, hessf, x0, epsilon, k=1000, plot=True)
              1.0 -
              0.5
              0.0 -
              -0.5
             -1.0
             -1.5
            -2.0 <del>↓</del>
-2
                           -1 0 1 2 0.00 0.25 0.50 0.75 1.00 |
| Iteration
              1.0
              -0.5
             -1.5
              -2.0 -
                                   0 1 2 0.00 0.25 0.50 0.75 1.00 Iteration
              1.5 -
              1.0 -
              0.5 -
              0.0 -
              -0.5
             -1.0
             -1.5
                                     0 1 2 0.00 0.25 0.50 0.75 1.00 Iteration
              2.0
              1.5 -
              0.5 -
              0.0 -
              -0.5
             -1.0
             -1.5
                                                        2 0.00 0.25 0.50 0.75 1.00
Iteration
```

For full credit on this problem, add some text description of what you observe about these results and why that is occurring (replace this text cell with your own thoughts).