

NotebookSolutions

September 1, 2024

```
[ ]: import numpy as np
import matplotlib.pyplot as plt
import plotly.graph_objects as go
```

1 Finding minimum value of a function

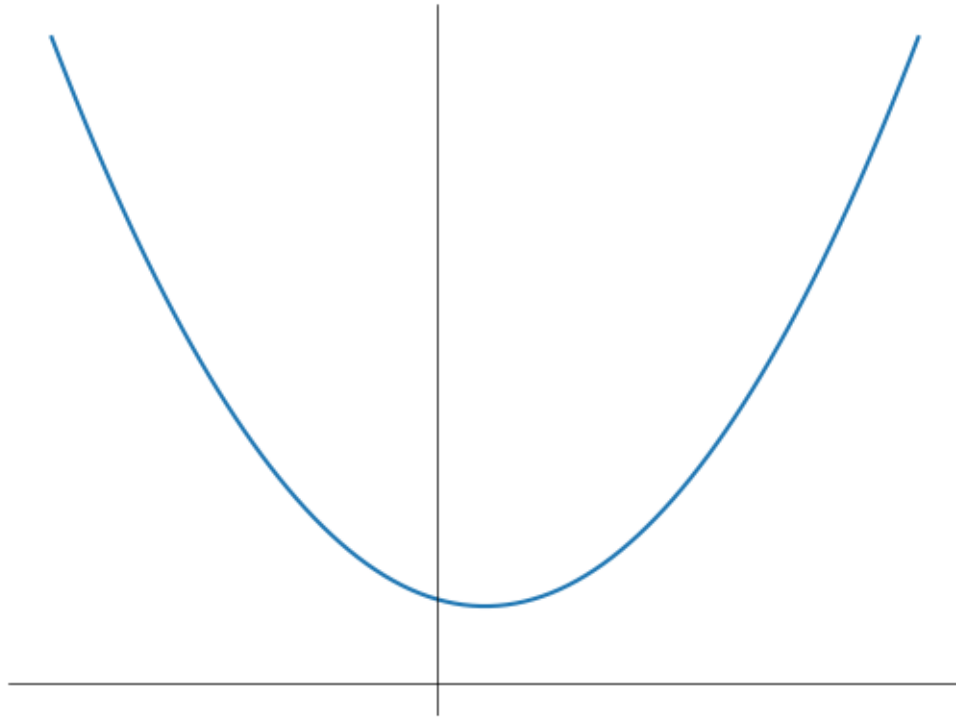
1.1 One dimensional function

$$f(x) = x^2 - x + 3$$

```
[ ]: def f(x):
    return x**2 - x + 3
```

Simple plot of the function. Note that if the function had many (> 2 parameters, then plotting would not be possible).

```
[ ]: xs = np.linspace(-4, 5, 100)
ys = f(xs)
plt.plot(xs, ys)
plt.axis('off')
plt.axhline(0, color='black', lw=0.5)
plt.axvline(0, color='black', lw=0.5)
plt.show()
```



1.1.1 Method 1: Grid search

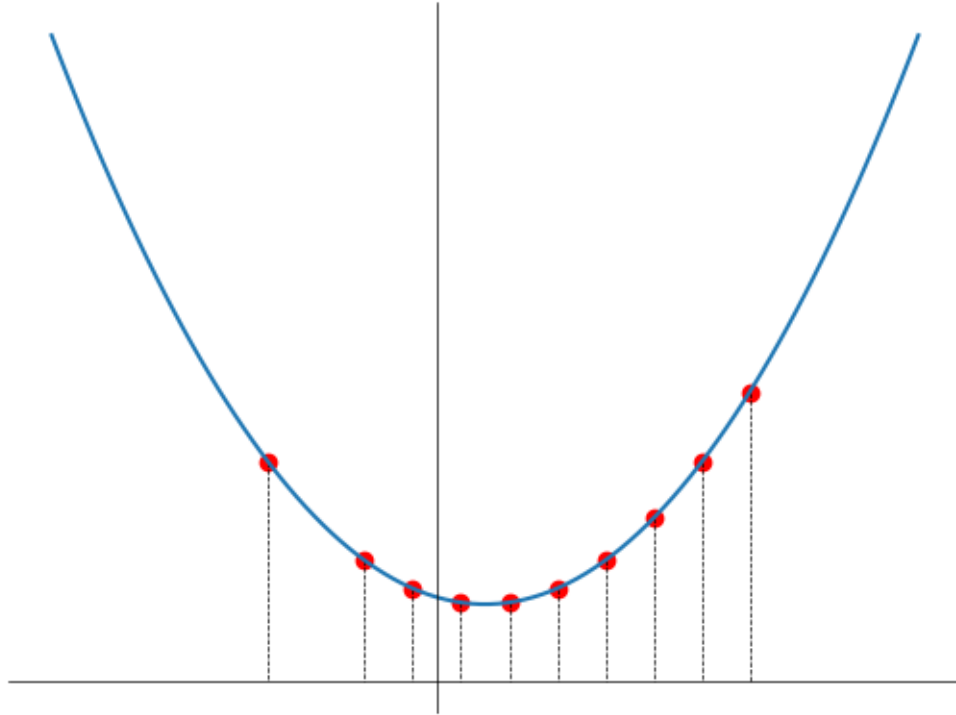
```
[ ]: x_vals = [-1.75, -0.75, -0.25, 0.25, 0.75, 1.25, 1.75, 2.25, 2.75, 3.25]
      for x in x_vals:
          print(f'f({x}) = {f(x)}')
```

```
f(-1.75) = 7.8125
f(-0.75) = 4.3125
f(-0.25) = 3.3125
f(0.25) = 2.8125
f(0.75) = 2.8125
f(1.25) = 3.3125
f(1.75) = 4.3125
f(2.25) = 5.8125
f(2.75) = 7.8125
f(3.25) = 10.3125
```

More visual version:

```
[ ]: x_vals = [-1.75, -0.75, -0.25, 0.25, 0.75, 1.25, 1.75, 2.25, 2.75, 3.25]
      for x in x_vals:
          plt.scatter(x, f(x), color='red')
          plt.plot([x, x], [f(x), 0], color='black', lw=0.5, linestyle='--')
      plt.plot(xs, ys)
```

```
plt.axis('off')
plt.axhline(0, color='black', lw=0.5)
plt.axvline(0, color='black', lw=0.5)
plt.show()
```



Minimum value of $f(x)$ found is 2.8125 at $x = 0.25$ and $x = 0.75$.

1.1.2 Method 2: Dichotomy

```
[ ]: a = -5
b = 5
for i in range(6):
    print(f"a: f({a}) = {f(a)} \t b: f({b}) = {f(b)}", end='\t\t')
    if f(a)>f(b):
        a = (a+b)/2
        print(f'set a = {a}')
    else:
        b = (a+b)/2
        print(f'set b = {b}')

x = ((a+b)/2)
print(f"best guess: x = {x}")
```

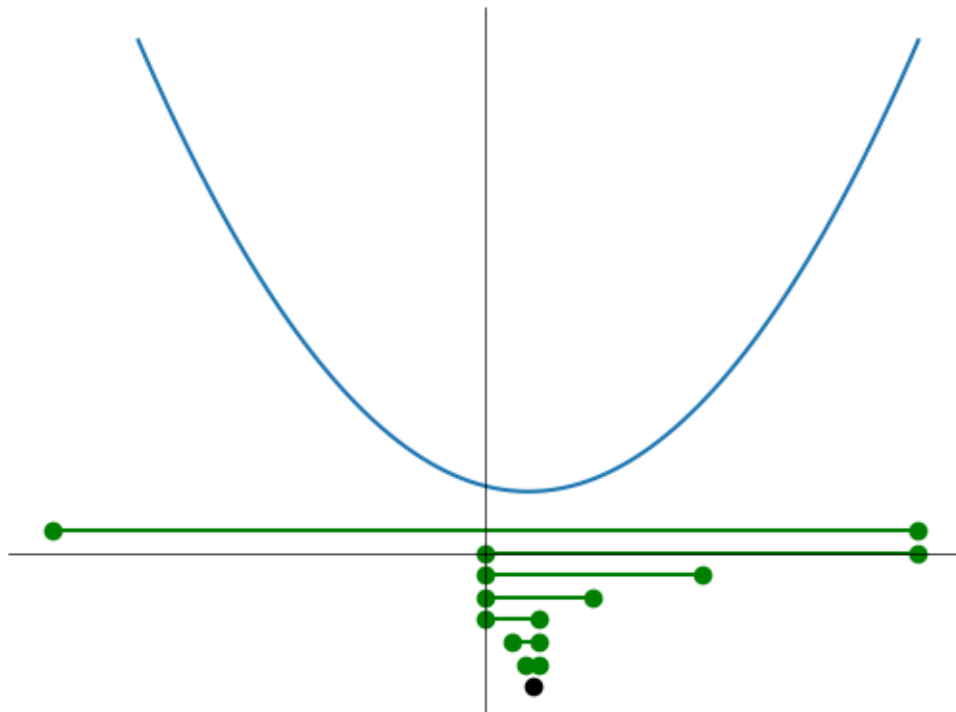
a: f(-5) = 33 b: f(5) = 23 set a = 0.0

a: f(0.0) = 3.0	b: f(5) = 23	set b = 2.5
a: f(0.0) = 3.0	b: f(2.5) = 6.75	set b = 1.25
a: f(0.0) = 3.0	b: f(1.25) = 3.3125	set b = 0.625
a: f(0.0) = 3.0	b: f(0.625) = 2.765625	set a = 0.3125
a: f(0.3125) = 2.78515625	b: f(0.625) = 2.765625	set a = 0.46875

best guess: x = 0.546875

More visual version

```
[ ]: a = -5
b = 5
plt.plot([a,b], [1, 1], color='green')
plt.scatter([a,b], [1, 1], color='green')
for i in range(6):
    if f(a)>f(b):
        a = (a+b)/2
    else:
        b = (a+b)/2
    plt.plot([a,b], [-i, -i], color='green')
    plt.scatter([a,b], [-i, -i], color='green')
x = ((a+b)/2)
plt.scatter([x], [-6], color='black')
plt.plot(xs, ys)
plt.axis('off')
plt.axhline(0, color='black', lw=0.5)
plt.axvline(0, color='black', lw=0.5)
plt.show()
```



1.1.3 Method 3: Gradient descent

We first calculate the derivative $f'(x) = 2x - 1$

```
[ ]: def df(x):  
      return 2*x-1
```

```
[ ]: lr = 0.8  
x = -1  
l = [x]  
for _ in range(7):  
    print(f"f({x:.2f}) = {f(x):.2f} \t f'({x:.2f}) = {df(x):.2f}")  
    x = x - lr * df(x)  
    l.append(x)  
print(f'Final guess: x = {x:.2f}')  
  
xs = np.linspace(min(l)-0.5, max(l+[2])+0.5, 100)  
ys = f(xs)
```

f(-1.00) = 5.00	f'(-1.00) = -3.00
f(1.40) = 3.56	f'(1.40) = 1.80
f(-0.04) = 3.04	f'(-0.04) = -1.08
f(0.82) = 2.85	f'(0.82) = 0.65
f(0.31) = 2.79	f'(0.31) = -0.39
f(0.62) = 2.76	f'(0.62) = 0.23
f(0.43) = 2.75	f'(0.43) = -0.14

Final guess: x = 0.54

Let's try other values for learning rate

```
[ ]: for lr in [0.3, 0.01, 1, 2]:  
    print(f'\t learning rate: {lr}')  
    x = -1  
    l = [x]  
    for _ in range(7):  
        print(f"f({x:.2f}) = {f(x):.2f} \t f'({x:.2f}) = {df(x):.2f}")  
        x = x - lr * df(x)  
        l.append(x)  
    print(f'Final guess: x = {x:.2f} \n\n\n')  
  
xs = np.linspace(min(l)-0.5, max(l+[2])+0.5, 100)  
ys = f(xs)
```

learning rate: 0.3	
f(-1.00) = 5.00	f'(-1.00) = -3.00
f(-0.10) = 3.11	f'(-0.10) = -1.20
f(0.26) = 2.81	f'(0.26) = -0.48

$f(0.40) = 2.76$	$f'(0.40) = -0.19$
$f(0.46) = 2.75$	$f'(0.46) = -0.08$
$f(0.48) = 2.75$	$f'(0.48) = -0.03$
$f(0.49) = 2.75$	$f'(0.49) = -0.01$

Final guess: $x = 0.50$

learning rate: 0.01

$f(-1.00) = 5.00$	$f'(-1.00) = -3.00$
$f(-0.97) = 4.91$	$f'(-0.97) = -2.94$
$f(-0.94) = 4.83$	$f'(-0.94) = -2.88$
$f(-0.91) = 4.74$	$f'(-0.91) = -2.82$
$f(-0.88) = 4.66$	$f'(-0.88) = -2.77$
$f(-0.86) = 4.59$	$f'(-0.86) = -2.71$
$f(-0.83) = 4.52$	$f'(-0.83) = -2.66$

Final guess: $x = -0.80$

learning rate: 1

$f(-1.00) = 5.00$	$f'(-1.00) = -3.00$
$f(2.00) = 5.00$	$f'(2.00) = 3.00$
$f(-1.00) = 5.00$	$f'(-1.00) = -3.00$
$f(2.00) = 5.00$	$f'(2.00) = 3.00$
$f(-1.00) = 5.00$	$f'(-1.00) = -3.00$
$f(2.00) = 5.00$	$f'(2.00) = 3.00$
$f(-1.00) = 5.00$	$f'(-1.00) = -3.00$

Final guess: $x = 2.00$

learning rate: 2

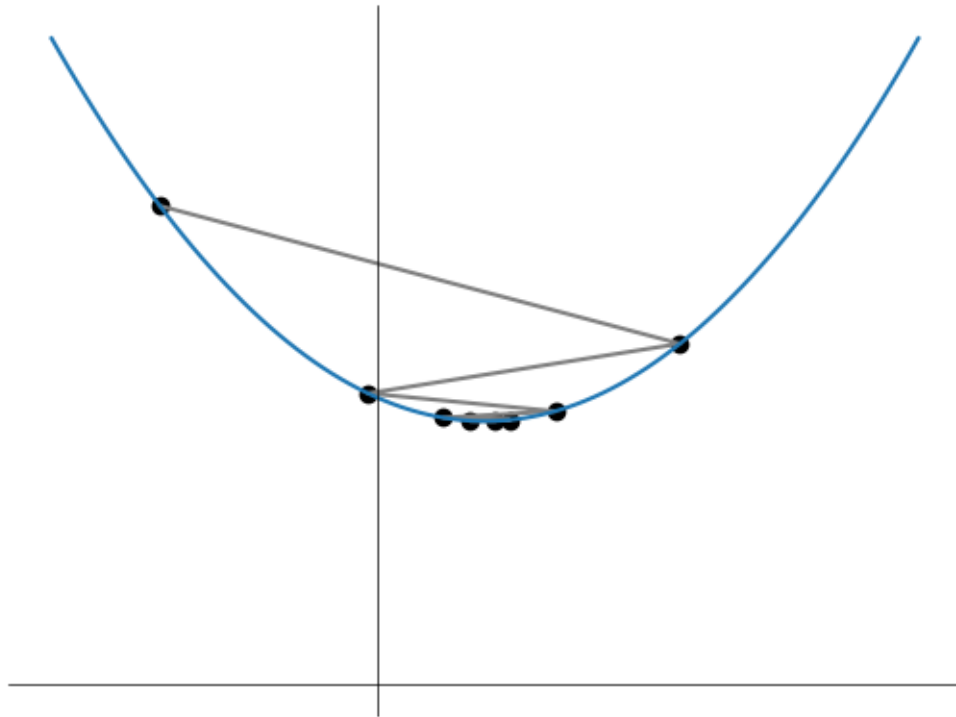
$f(-1.00) = 5.00$	$f'(-1.00) = -3.00$
$f(5.00) = 23.00$	$f'(5.00) = 9.00$
$f(-13.00) = 185.00$	$f'(-13.00) = -27.00$
$f(41.00) = 1643.00$	$f'(41.00) = 81.00$
$f(-121.00) = 14765.00$	$f'(-121.00) = -243.00$
$f(365.00) = 132863.00$	$f'(365.00) = 729.00$
$f(-1093.00) = 1195745.00$	$f'(-1093.00) = -2187.00$

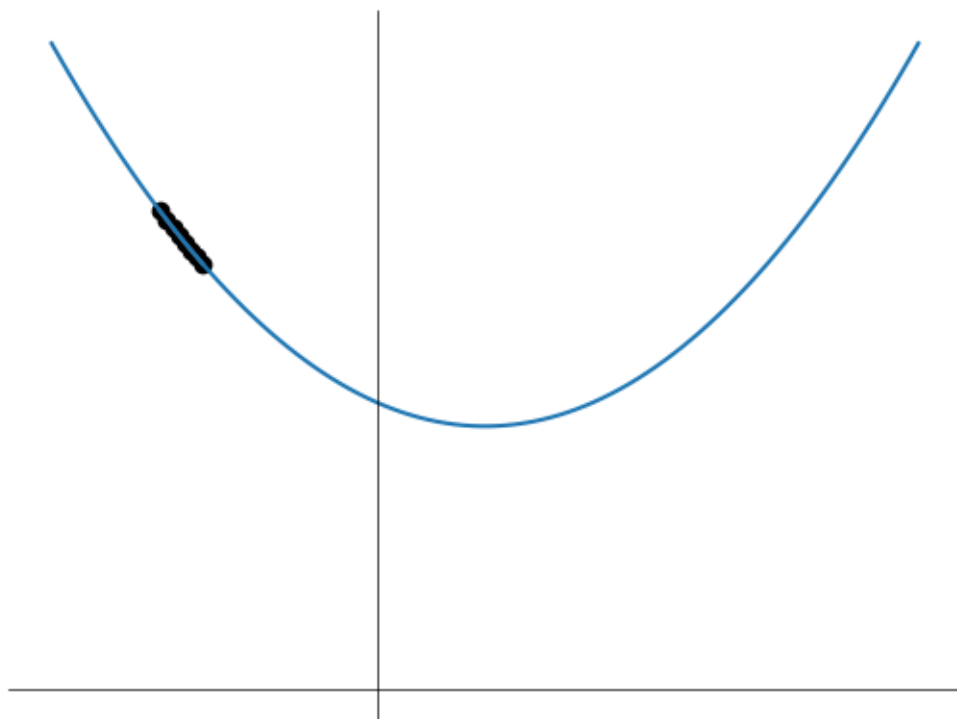
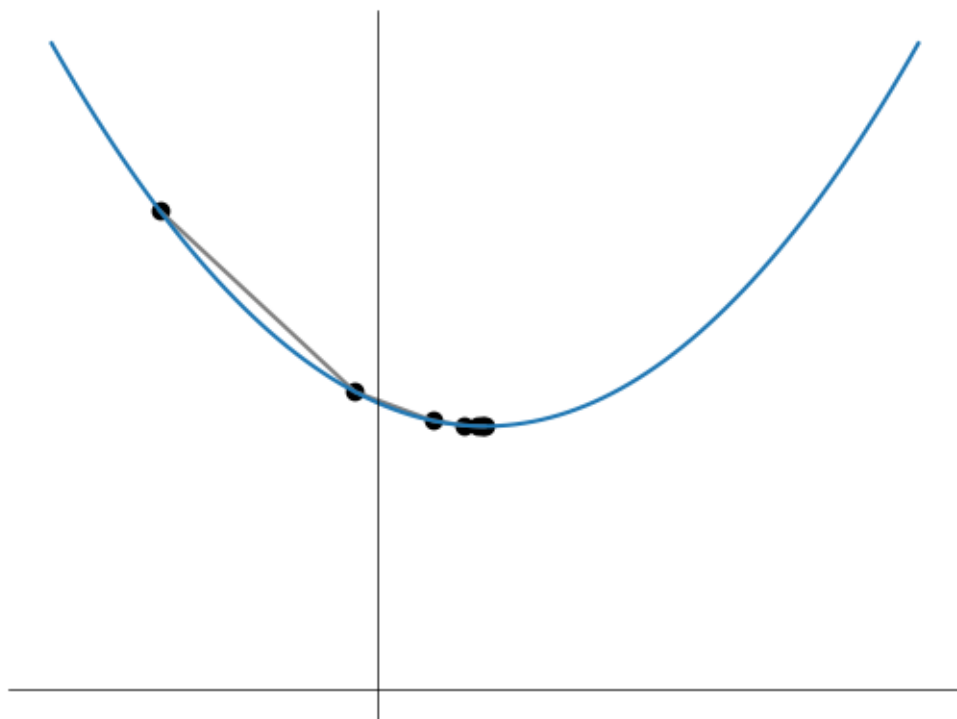
Final guess: $x = 3281.00$

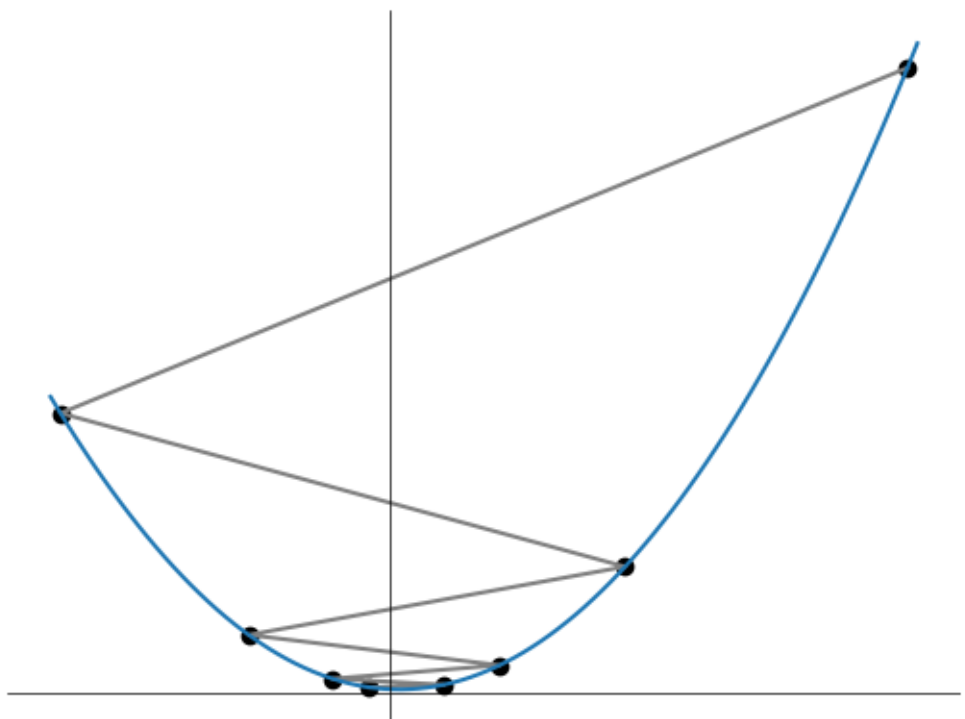
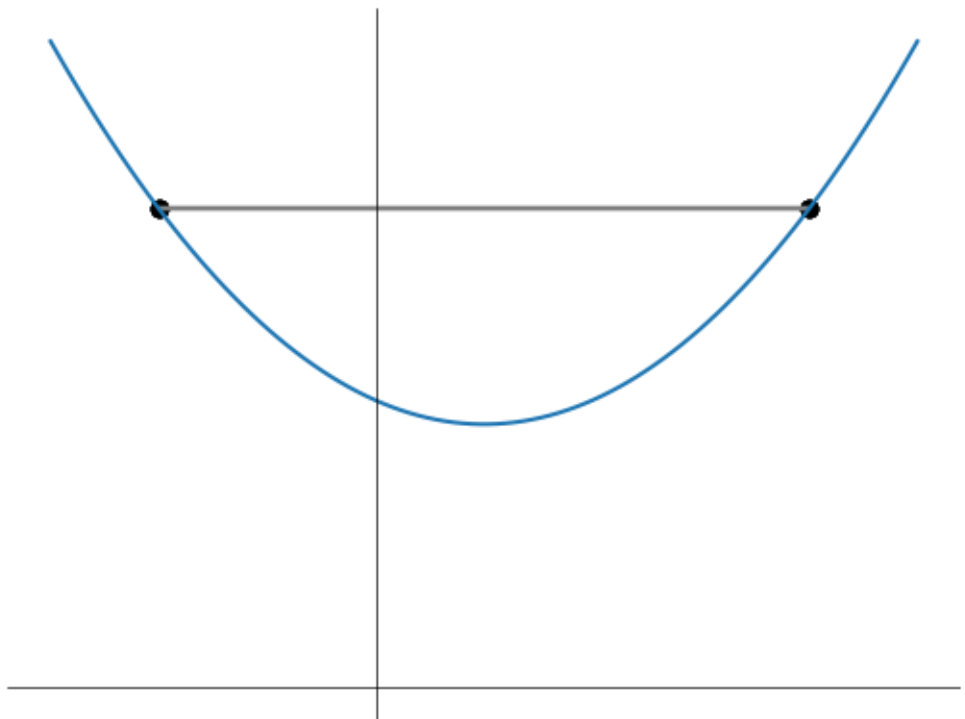
Visually:

```
[ ]: for lr in [0.8, 0.3, 0.01, 1, 1.25]:
    x = -1
    l = [x]
    for _ in range(7):
        x = x - lr * df(x)
        l.append(x)

    plt.plot(l, [f(x) for x in l], color='grey')
    plt.scatter(l, [f(x) for x in l], color='black')
    xs = np.linspace(min(l)-0.5, max(l+[2])+0.5, 100)
    ys = f(xs)
    plt.plot(xs, ys)
    plt.axis('off')
    plt.axhline(0, color='black', lw=0.5)
    plt.axvline(0, color='black', lw=0.5)
    plt.show()
```







1.2 Two dimensional function

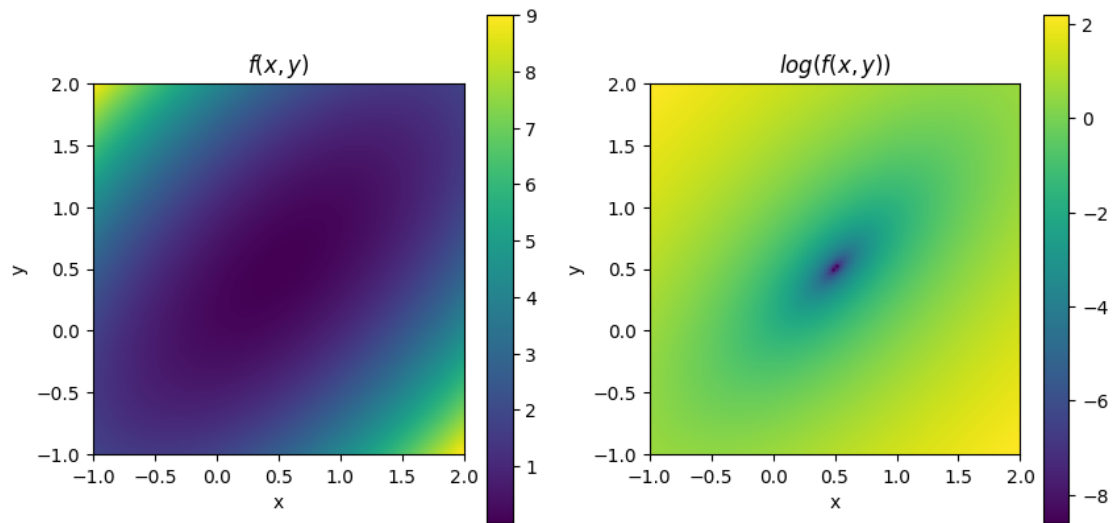
$$f(x, y) = (x + y - 1)^2 + \frac{1}{5}(x - y)^2$$

```
[ ]: def f(x, y):  
      return (x+y-1)**2 + 0.2*(x-y)**2
```

Plotting the function as a 2D surface. Plotting log of the function helps to see the minimum more clearly.

```
[ ]: # interactive  
x = np.linspace(-1, 2, 100)  
y = np.linspace(-1, 2, 100)  
X, Y = np.meshgrid(x, y)  
Z = f(X, Y)  
Zlog = np.log(Z)  
go.Figure(data=[go.Surface(z=Z)])
```

```
[ ]: # not interactive  
x = np.linspace(-1, 2, 100)  
y = np.linspace(-1, 2, 100)  
X, Y = np.meshgrid(x, y)  
Z = f(X, Y)  
Zlog = np.log(Z)  
  
# plot a heatmap of the function  
plt.figure(figsize=(10,5))  
plt.subplot(1,2,1)  
plt.imshow(Z, extent=(-1, 2, -1, 2))  
plt.colorbar()  
plt.title('$f(x,y)$')  
plt.xlabel('x')  
plt.ylabel('y')  
plt.subplot(1,2,2)  
plt.imshow(Zlog, extent=(-1, 2, -1, 2))  
plt.colorbar()  
plt.title('$\log(f(x,y))$')  
plt.xlabel('x')  
plt.ylabel('y')  
plt.show()
```



1.2.1 Method 1: Grid search

```
[ ]: for x in np.arange(-1, 2.5, 0.5):
      for y in np.arange(-1, 2.5, 0.5):
          print(f'f({x}, {y}) = {f(x, y)}')
```

```
f(-1.0, -1.0) = 9.0
f(-1.0, -0.5) = 6.3
f(-1.0, 0.0) = 4.2
f(-1.0, 0.5) = 2.7
f(-1.0, 1.0) = 1.8
f(-1.0, 1.5) = 1.5
f(-1.0, 2.0) = 1.8
f(-0.5, -1.0) = 6.3
f(-0.5, -0.5) = 4.0
f(-0.5, 0.0) = 2.3
f(-0.5, 0.5) = 1.2
f(-0.5, 1.0) = 0.7
f(-0.5, 1.5) = 0.8
f(-0.5, 2.0) = 1.5
f(0.0, -1.0) = 4.2
f(0.0, -0.5) = 2.3
f(0.0, 0.0) = 1.0
f(0.0, 0.5) = 0.3
f(0.0, 1.0) = 0.2
f(0.0, 1.5) = 0.7
f(0.0, 2.0) = 1.8
f(0.5, -1.0) = 2.7
f(0.5, -0.5) = 1.2
f(0.5, 0.0) = 0.3
```

```

f(0.5, 0.5) = 0.0
f(0.5, 1.0) = 0.3
f(0.5, 1.5) = 1.2
f(0.5, 2.0) = 2.7
f(1.0, -1.0) = 1.8
f(1.0, -0.5) = 0.7
f(1.0, 0.0) = 0.2
f(1.0, 0.5) = 0.3
f(1.0, 1.0) = 1.0
f(1.0, 1.5) = 2.3
f(1.0, 2.0) = 4.2
f(1.5, -1.0) = 1.5
f(1.5, -0.5) = 0.8
f(1.5, 0.0) = 0.7
f(1.5, 0.5) = 1.2
f(1.5, 1.0) = 2.3
f(1.5, 1.5) = 4.0
f(1.5, 2.0) = 6.3
f(2.0, -1.0) = 1.8
f(2.0, -0.5) = 1.5
f(2.0, 0.0) = 1.8
f(2.0, 0.5) = 2.7
f(2.0, 1.0) = 4.2
f(2.0, 1.5) = 6.3
f(2.0, 2.0) = 9.0

```

Note that it becomes computationally expensive to search for the minimum in a 2D grid. For higher dimensions, this method is not feasible.

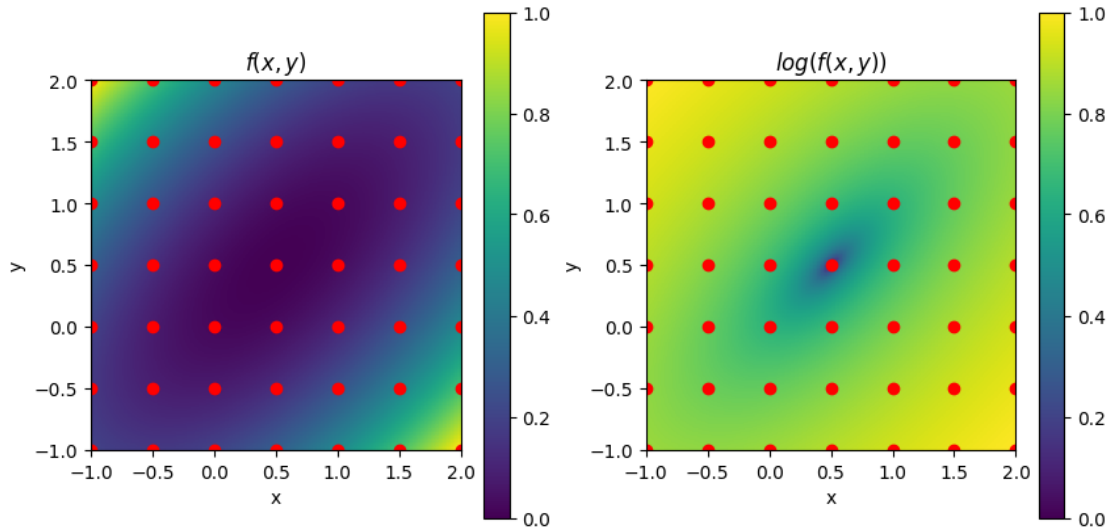
More visually:

```

[ ]: # plot a heatmap of the function
plt.figure(figsize=(10,5))
plt.subplot(1,2,1)
plt.imshow(Z, extent=(-1, 2, -1, 2))
for x in np.arange(-1, 2.5, 0.5):
    for y in np.arange(-1, 2.5, 0.5):
        plt.scatter(x, y, color='red')
plt.colorbar()
plt.title('$f(x,y)$')
plt.xlabel('x')
plt.ylabel('y')
plt.subplot(1,2,2)
plt.imshow(Zlog, extent=(-1, 2, -1, 2))
for x in np.arange(-1, 2.5, 0.5):
    for y in np.arange(-1, 2.5, 0.5):
        plt.scatter(x, y, color='red')
plt.colorbar()

```

```
plt.title('$\log(f(x,y))$')
plt.xlabel('x')
plt.ylabel('y')
plt.show()
```



1.2.2 Method 2: Dichotomy

```
[ ]: a, b = -10, 10
      c, d = -10, 10
      for _ in range(10):
          f1 = f(a, c)
          f2 = f(a, d)
          f3 = f(b, c)
          f4 = f(b, d)
          if f1<=f2 and f1<=f3 and f1<=f4:
              b = (a+b)/2
              d = (c+d)/2
          if f2<=f1 and f2<=f3 and f2<=f4:
              b = (a+b)/2
              c = (c+d)/2
          if f3<=f2 and f3<=f1 and f3<=f4:
              a = (a+b)/2
              d = (c+d)/2
          if f4<=f2 and f4<=f3 and f4<=f1:
              a = (a+b)/2
              c = (c+d)/2
          print(f'f({(a+b)/2:.2f}, {(c+d)/2:.2f}) = {f((a+b)/2, (c+d)/2):.4f} \t {a:.2f} < x < {b:.2f} \t {c:.2f} < y < {d:.2f} \t ')
          print('')
```

```
print(f'Best guess: x = {(a+b)/2:.2f}, y = {(c+d)/2:.2f}')
```

$f(-2.50, 2.50) = 6.0000$	$-5.00 < x < 0.00$	$0.00 < y < 5.00$
$f(-1.25, 1.25) = 2.2500$	$-2.50 < x < 0.00$	$0.00 < y < 2.50$
$f(-0.62, 0.62) = 1.3125$	$-1.25 < x < 0.00$	$0.00 < y < 1.25$
$f(-0.31, 0.94) = 0.4531$	$-0.62 < x < 0.00$	$0.62 < y < 1.25$
$f(-0.16, 0.78) = 0.3164$	$-0.31 < x < 0.00$	$0.62 < y < 0.94$
$f(-0.08, 0.86) = 0.2236$	$-0.16 < x < 0.00$	$0.78 < y < 0.94$
$f(-0.04, 0.82) = 0.1956$	$-0.08 < x < 0.00$	$0.78 < y < 0.86$
$f(-0.02, 0.84) = 0.1800$	$-0.04 < x < 0.00$	$0.82 < y < 0.86$
$f(-0.01, 0.83) = 0.1734$	$-0.02 < x < 0.00$	$0.82 < y < 0.84$
$f(-0.00, 0.83) = 0.1699$	$-0.01 < x < 0.00$	$0.83 < y < 0.84$

Best guess: $x = -0.00$, $y = 0.83$

More visual version

```
[ ]: a, b = -5, 5
      c, d = -5, 5

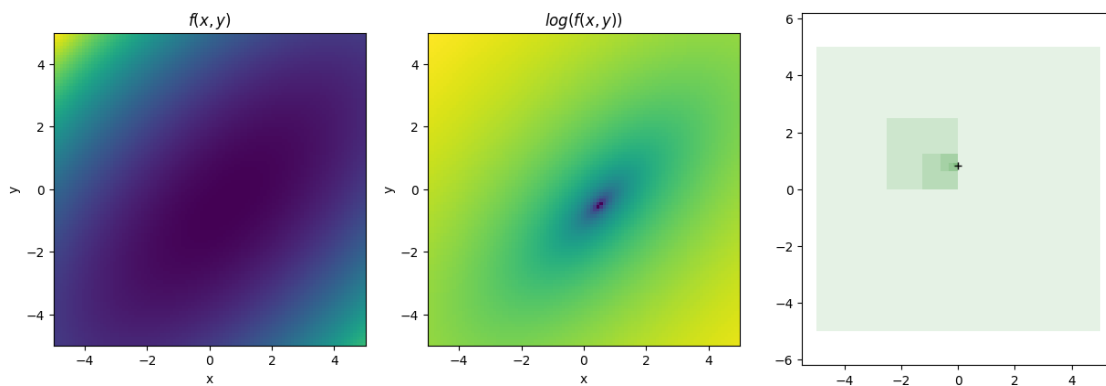
      fig = plt.figure(figsize=(15,5))
      x = np.linspace(a,b, 100)
      y = np.linspace(c,d, 100)
      X, Y = np.meshgrid(x, y)
      Z = f(X, Y)
      Zlog = np.log(Z)
      plt.subplot(1,3,1)
      plt.imshow(Z, extent=(a, b, c, d))
      plt.title('$f(x,y)$')
      plt.xlabel('x')
      plt.ylabel('y')
      plt.subplot(1,3,2)
      plt.imshow(Zlog, extent=(a, b, c, d))
      plt.title('$\log(f(x,y))$')
      plt.xlabel('x')
      plt.ylabel('y')

      ax = plt.subplot(1,3,3)
      rectangles = []
      for _ in range(10):
          rectangle = plt.Rectangle((a, c), b-a, d-c, facecolor='green', alpha=0.1)
          ax.add_patch(rectangle)
          f1 = f(a, c)
          f2 = f(a, d)
          f3 = f(b, c)
          f4 = f(b, d)
          if f1<=f2 and f1<=f3 and f1<=f4:
              b = (a+b)/2
```

```

    d = (c+d)/2
    if f2<=f1 and f2<=f3 and f2<=f4:
        b = (a+b)/2
        c = (c+d)/2
    if f3<=f2 and f3<=f1 and f3<=f4:
        a = (a+b)/2
        d = (c+d)/2
    if f4<=f2 and f4<=f3 and f4<=f1:
        a = (a+b)/2
        c = (c+d)/2
plt.plot((a+b)/2, (c+d)/2, '+', color='black')
plt.axis('equal')
plt.show()

```



1.2.3 Method 3: Gradient descent

We start by calculating the partial derivatives of the function with respect to x and y .

$$\frac{\partial f}{\partial x}(x, y) = 2(x + y - 1) + \frac{2}{5}(x - y)$$

$$\frac{\partial f}{\partial y}(x, y) = 2(x + y - 1) - \frac{2}{5}(x - y)$$

```

[ ]: def dfx(x, y):
    return 2*(x+y-1) + 0.4*(x-y)
def dfy(x, y):
    return 2*(x+y-1) - 0.4*(x-y)

```

```

[ ]: lr = 0.6
x = 2.5
y = 3
lx = [x]
ly = [y]
for _ in range(7):
    x = x - lr * dfx(x,y)

```

```

y = y - lr * dfy(x,y)
lx.append(x)
ly.append(y)
print(f'f(x,y) = {f(x,y):.2f} \t (x,y) = ({x:.2f}, {y:.2f})')

```

```

f(x,y) = 7.20      (x,y) = (-2.78, 2.55)
f(x,y) = 0.85      (x,y) = (-0.02, 0.10)
f(x,y) = 0.25      (x,y) = (1.11, 0.09)
f(x,y) = 0.04      (x,y) = (0.63, 0.56)
f(x,y) = 0.01      (x,y) = (0.39, 0.58)
f(x,y) = 0.00      (x,y) = (0.47, 0.49)
f(x,y) = 0.00      (x,y) = (0.52, 0.48)

```

```

[ ]: lr = 0.6
for lr in [0.3, 0.1, 0.01, 1.25]:
    print(f'learning rate: {lr}')
    x = 2.5
    y = 3
    lx = [x]
    ly = [y]
    for _ in range(7):
        x = x - lr * dfx(x,y)
        y = y - lr * dfy(x,y)
        lx.append(x)
        ly.append(y)
        print(f'f(x,y) = {f(x,y):.2f} \t (x,y) = ({x:.2f}, {y:.2f})')
    print('\n\n\n')

```

```

learning rate: 0.3
f(x,y) = 0.68      (x,y) = (-0.14, 1.51)
f(x,y) = 0.32      (x,y) = (-0.16, 1.10)
f(x,y) = 0.16      (x,y) = (0.03, 0.90)
f(x,y) = 0.07      (x,y) = (0.18, 0.77)
f(x,y) = 0.03      (x,y) = (0.28, 0.68)
f(x,y) = 0.01      (x,y) = (0.35, 0.62)
f(x,y) = 0.01      (x,y) = (0.40, 0.58)

```

```

learning rate: 0.1
f(x,y) = 8.14      (x,y) = (1.62, 2.22)
f(x,y) = 3.29      (x,y) = (1.08, 1.72)
f(x,y) = 1.36      (x,y) = (0.74, 1.39)
f(x,y) = 0.58      (x,y) = (0.54, 1.17)
f(x,y) = 0.27      (x,y) = (0.43, 1.02)
f(x,y) = 0.14      (x,y) = (0.36, 0.92)
f(x,y) = 0.08      (x,y) = (0.33, 0.84)

```



```

learning rate: 0.01
f(x,y) = 18.72    (x,y) = (2.41, 2.91)
f(x,y) = 17.27    (x,y) = (2.33, 2.82)
f(x,y) = 15.93    (x,y) = (2.25, 2.74)
f(x,y) = 14.69    (x,y) = (2.17, 2.66)
f(x,y) = 13.55    (x,y) = (2.09, 2.58)
f(x,y) = 12.50    (x,y) = (2.02, 2.51)
f(x,y) = 11.53    (x,y) = (1.95, 2.44)

```

```

learning rate: 1.25
f(x,y) = 112.80      (x,y) = (-8.50, 13.50)
f(x,y) = 324.80      (x,y) = (-7.50, -9.50)
f(x,y) = 1804.80     (x,y) = (36.50, -51.50)
f(x,y) = 5196.80     (x,y) = (32.50, 40.50)
f(x,y) = 28876.80    (x,y) = (-143.50, 208.50)
f(x,y) = 83148.80    (x,y) = (-127.50, -159.50)
f(x,y) = 462028.80   (x,y) = (576.50, -831.50)

```

A more visual version of the gradient descent method

```

[ ]: for lr in [0.6, 0.3, 0.1, 0.01, 1.25]:
    x = 2.5
    y = 3
    lx = [x]
    ly = [y]
    for _ in range(7):
        x = x - lr * dfx(x,y)
        y = y - lr * dfy(x,y)
        lx.append(x)
        ly.append(y)

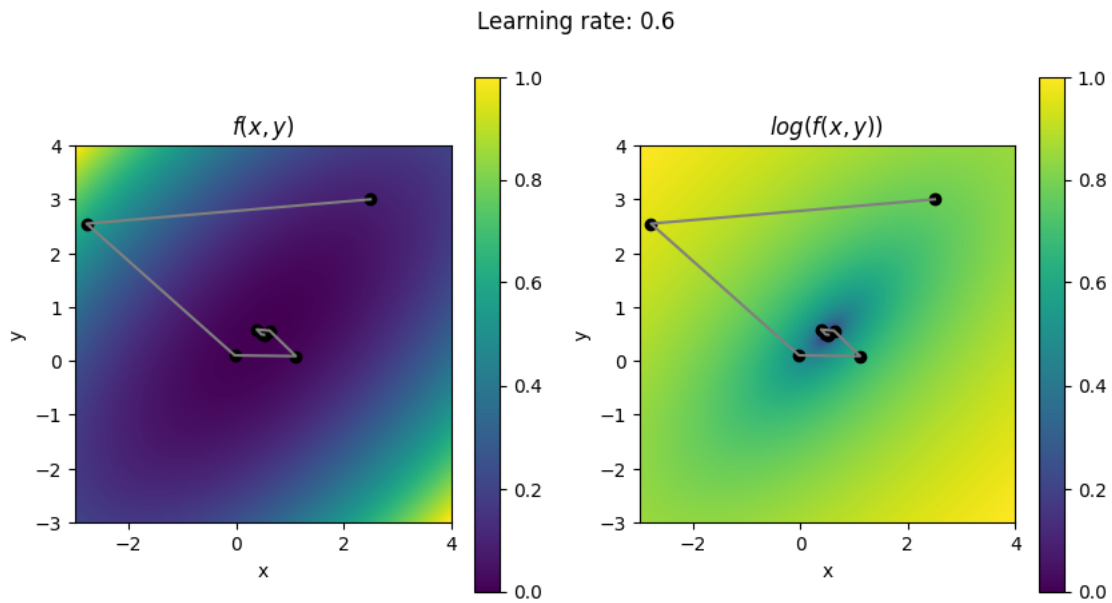
    # plot a heatmap of the function with the line search
    x = np.linspace(-3, 4, 100)
    y = np.linspace(-3, 4, 100)
    X, Y = np.meshgrid(x, y)
    Z = f(X, Y)
    Zlog = np.log(Z)

```

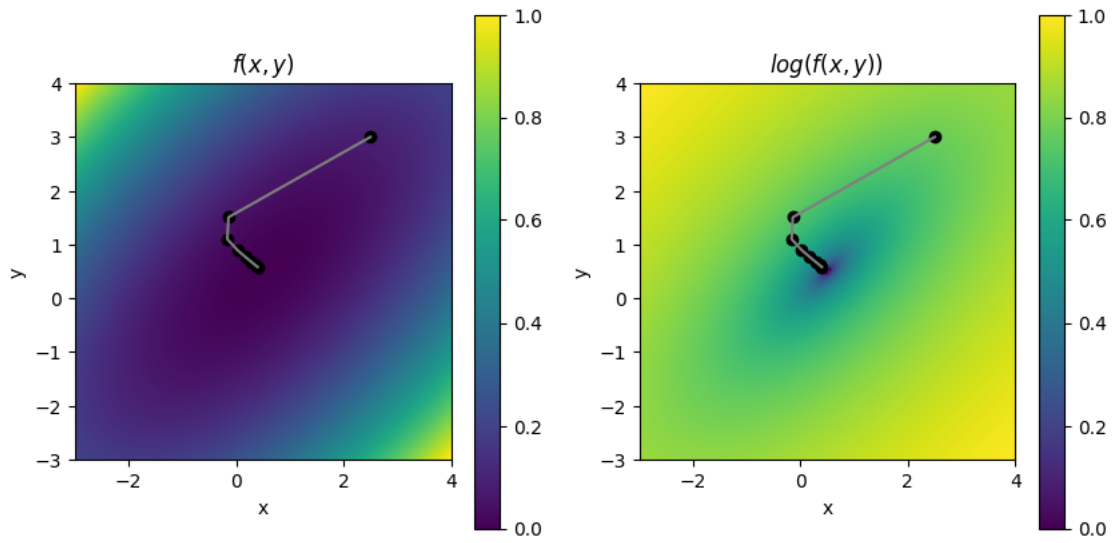
```

plt.figure(figsize=(10,5))
plt.subplot(1,2,1)
plt.imshow(Z, extent=(-3, 4, -3, 4))
plt.plot(lx, ly, color='grey')
plt.scatter(lx, ly, color='black')
plt.colorbar()
plt.title('$f(x,y)$')
plt.xlabel('x')
plt.ylabel('y')
plt.subplot(1,2,2)
plt.imshow(Zlog, extent=(-3, 4, -3, 4))
plt.plot(lx, ly, color='grey')
plt.scatter(lx, ly, color='black')
plt.colorbar()
plt.title('$\log(f(x,y))$')
plt.xlabel('x')
plt.ylabel('y')
plt.suptitle(f'Learning rate: {lr}')
plt.show()

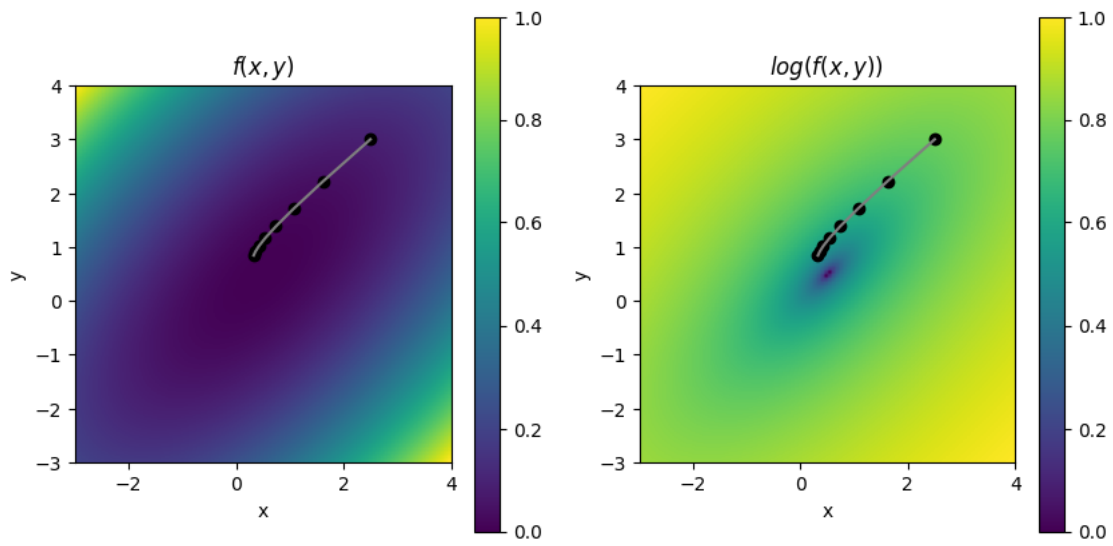
```



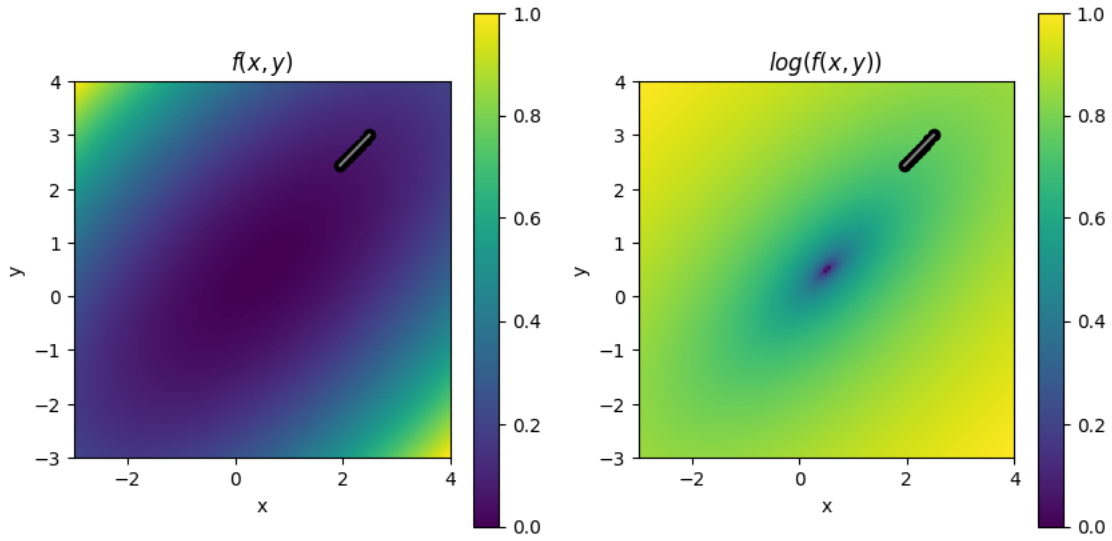
Learning rate: 0.3



Learning rate: 0.1



Learning rate: 0.01



Learning rate: 1.25

