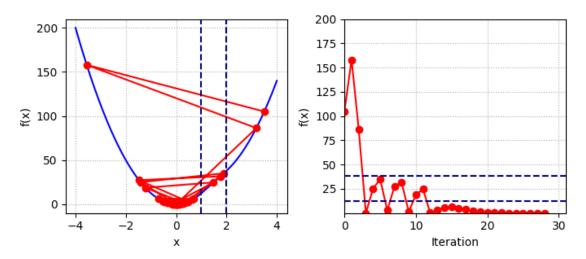
March 24, 2024

```
[1]: import numpy as np
     import matplotlib.pyplot as plt
     import matplotlib.animation as animation
     from IPython.display import HTML
[2]: def grad_f(x):
         if x < 1:
             return 25 * x
         if 1 \le x \le 2:
             return x + 24
         return 25 * x - 24
     def heavy_ball_method(alpha, beta, x0, num_iterations):
         x = np.zeros(num_iterations + 1)
         x_prev = x0
         x_curr = x0
         for i in range(num_iterations):
             x[i] = x_{curr}
             x_new = x_curr - alpha * grad_f(x_curr) + beta * (x_curr - x_prev)
             x_prev = x_curr
             x_curr = x_new
         x[num_iterations] = x_curr
         return x
     L = 25 # directly from gradient
     mu = 1.0 # from second order differential criterion
     alpha_star = 4 / (np.sqrt(L) + np.sqrt(mu))**2
     beta_star = (np.sqrt(L) - np.sqrt(mu))**2 / (np.sqrt(L) + np.sqrt(mu))**2
     x0 = 3.5
     num_iterations = 30
     trajectory = heavy_ball_method(alpha_star, beta_star, x0, num_iterations)
     fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(7, 3.5))
     fig.suptitle("Heavy ball method with optimal hyperparameters")
     def update(i):
         ax1.clear()
```

```
ax2.clear()
    x_vals = np.linspace(-4, 4, 100)
    f_{vals} = np.piecewise(x_{vals}, [x_{vals} < 1, (x_{vals} >= 1) & (x_{vals} < 2),
 \rightarrowx_vals >= 2],
                         [lambda x: 12.5 * x**2, lambda x: .5 * x**2 + 24 * x - 11
\rightarrow12, lambda x: 12.5 * x**2 - 24 * x + 36])
    ax1.plot(x_vals, f_vals, 'b-')
    ax1.plot(trajectory[:i], [12.5 * x**2 if x < 1 else .5 * x**2 + 24 * x - 12_{u}
\rightarrow if x < 2 else 12.5 * x**2 - 24 * x + 36 for x in trajectory[:i]], 'ro-')
    ax1.axvline(x=1, color='navy', linestyle='--')
    ax1.axvline(x=2, color='navy', linestyle='--')
    f_trajectory = [None for x in trajectory]
    f_{trajectory}[:i] = [12.5 * x**2 if x < 1 else .5 * x**2 + 24 * x - 12 if x < 1]
\rightarrow2 else 12.5 * x**2 - 24 * x + 36 for x in trajectory[:i]]
    ax2.plot(range(len(trajectory)), f_trajectory, 'ro-')
    ax2.set_xlim(0, len(trajectory))
    ax2.set_ylim(min(f_vals), max(f_vals))
    f_1 = 12.5 * 1.0**2
    f_2 = .5 * 2.**2 + 24 * 2. - 12
    ax2.axhline(y=f_1, color='navy', linestyle='--')
    ax2.axhline(y=f_2, color='navy', linestyle='--')
    ax1.set_xlabel("x")
    ax1.set_ylabel("f(x)")
    ax1.grid(linestyle=":")
    ax2.set_xlabel("Iteration")
    ax2.set_ylabel("f(x)")
    ax2.grid(linestyle=":")
    plt.tight_layout()
for i in range(num_iterations):
    update(i)
```

Heavy ball method with optimal hyperparameters $\alpha^* \beta^*$



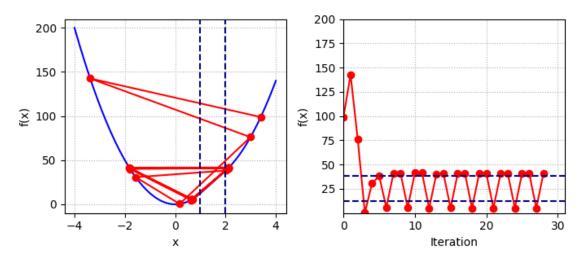
```
[3]: L = 25  # directly from gradient
    mu = 1.0  # from second order differential criterion
    alpha_star = 4 / (np.sqrt(L) + np.sqrt(mu))**2
    beta_star = (np.sqrt(L) - np.sqrt(mu))**2 / (np.sqrt(L) + np.sqrt(mu))**2
    x0 = 3.4
    num_iterations = 30

    trajectory = heavy_ball_method(alpha_star, beta_star, x0, num_iterations)

fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(7, 3.5))
    fig.suptitle("Heavy ball method with optimal hyperparameters")

for i in range(num_iterations):
    update(i)
```

Heavy ball method with optimal hyperparameters $\alpha^* \beta^*$



```
[4]: L = 25  # directly from gradient
    mu = 1.0  # from second order differential criterion
    alpha_star = 2 / L
    beta_star = mu / L
    x0 = 3.4
    num_iterations = 30

    trajectory = heavy_ball_method(alpha_star, beta_star, x0, num_iterations)

fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(7, 3.5))
    fig.suptitle("Heavy ball method with optimal hyperparameters")

for i in range(num_iterations):
    update(i)
```

Heavy ball method with optimal hyperparameters α^* β^*

