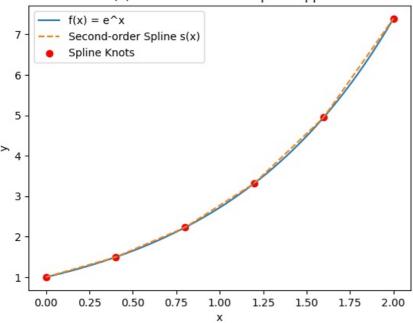
```
In [11]: #Code implementation (your own, no existing method for interpolation).
         import matplotlib.pyplot as plt
         import numpy as np
         import pandas as pd
In [12]: \#f(x), a,b, h,N x_i,y_i
         f = np.exp
         a, b = 0, 2
         N = 5
         h = (b - a) / N
         x_{points} = np.linspace(a, b, N + 1)
         f_values = f(x_points)
In [13]: # Second-order spline coefficients calculation
         coeffs = []
         for i in range(len(x points) - 1):
             x_i, x_next = x_points[i], x_points[i + 1]
             f_i, f_next = f_values[i], f_values[i + 1]
             # Calculate coefficients for the quadratic spline
             a_i = f_i
             b i = (f next - f i) / h # Using h directly
             c_i = (f_next - f_i - b_i * h) / (h ** 2) # Using h in the quadratic term
             coeffs.append((a i, b i, c i))
In [14]: # Test points to evaluate the spline
         x_{test} = np.linspace(a, b, 50)
         f_{\text{test}} = f(x_{\text{test}})
         # Calculate spline values s(x) and delta at test points
         s values = []
         \overline{deltas} = []
         for x in x_test:
             for i in range(len(x points) - 1):
                 if x_points[i] <= x <= x_points[i + 1]:
                      a_i, b_i, c_i = coeffs[i]
                     x_i = x_points[i]
                      s_x = a_i + b_i * (x - x_i) + c_i * (x - x_i) ** 2
                      s_values.append(s_x)
                      deltas.append(abs(f(x) - s_x))
                     break
In [15]: # Output of the code in form of the table:
         results_df = pd.DataFrame({
             "x": x_test,
             "f(x)": f_test,
             "s(x)": s values,
             "delta": deltas
         })
         results_df.head()
                                          delta
                                  s(x)
         0 0.000000 1.000000 1.000000 0.000000
         1 0.040816 1.041661 1.050186 0.008525
         2 0.081633 1.085057 1.100372 0.015315
         3 0.122449 1.130261 1.150559 0.020297
         4 0.163265 1.177349 1.200745 0.023396
In [16]: #Graphs:
         #Graph of distribution of real function f(x) and spline of the first order and the second order (s 1(x) and s 2
         #Dependence of mean average error value on number of points used in your domain.
         plt.figure()
         plt.plot(x_test, f_test, label="f(x) = e^x")
         plt.plot(x_test, s_values, label="Second-order Spline s(x)", linestyle="--")
         plt.scatter(x points, f values, color="red", marker="o", label="Spline Knots")
         plt.legend()
         plt.xlabel("x")
         plt.ylabel("y")
         plt.title("Function f(x) and Second-order Spline Approximation")
         plt.show()
```

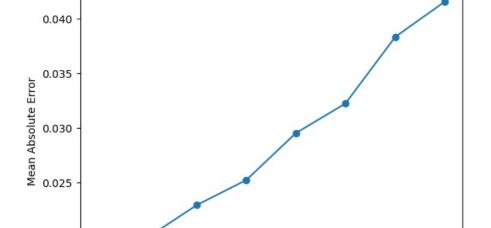
Function f(x) and Second-order Spline Approximation



```
In [17]: # Mean absolute error vs. number of points
points_range = range(3, 11)
    mean_errors = [np.mean(deltas[:int(len(deltas) * p / max(points_range))]) for p in points_range]

plt.figure()
plt.plot(list(points_range), mean_errors, marker="o")
plt.xlabel("Number of Points")
plt.ylabel("Mean Absolute Error")
plt.title("Mean Absolute Error vs. Number of Points")
plt.show()
```

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Number of Points

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Mean Absolute Error vs. Number of Points

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