1 Интерполирование с равноудалёнными узлами

```
[269]: import matplotlib.pyplot as plt
       import numpy as np
       from numpy import sinh
       from IPython.display import display, Latex
       def newton_polynomial(x: float, nodes_x: list, nodes_f: list) -> float:
           if len(nodes_x) != len(nodes_f):
               return None
           h = nodes_x[1] - nodes_x[0] # step between equidistant interpolation nodes
           polynomial = 0
           if x < nodes_x[round(len(nodes_x) / 1.5)]: # checks if x is at the_
        →beginning of the table
               t = (x - nodes_x[0]) / h
               factor = 1
               for i in range(len(nodes_x)):
                   polynomial += finite_differences(0, nodes_f, order=i) * factor
                   factor *= (t - i) / (i + 1)
           elif x < nodes_x[round(len(nodes_x) / 1.5)]: # checks if x is at the center_
        \rightarrow of the table
               n = 0
               for i in range(int(len(nodes_x) / 3), len(nodes_x)): # finding n - - 
        →closest node's number
                   if nodes_x[i] > x:
                       n = i \text{ if } nodes_x[i] - x < x - nodes_x[i - 1] \text{ else } i - 1
                       break
               t = (x - nodes x[n]) / h
               factor = 1
               for i in range(len(nodes_x)):
                   k = round(i / 2)
                   polynomial += finite_differences(n - k, nodes_f, order=i) * factor
                   factor *= t + k if i \% 2 == 0 else t - k
                   factor /= i + 1
           else: # x is at the end of the table
               t = (x - nodes_x[-1]) / h
               n = len(nodes_x)
               factor = 1
               for i in range(n):
                   polynomial += finite_differences(n - i - 1, nodes_f, order=i) *__
        \rightarrowfactor
                   factor *= (t + i) / (i + 1)
           return polynomial
```

```
[270]: x_nodes = tuple(x / 10 for x in range(10,19))
      f_nodes = (1.17520, 1.33565, 1.50946, 1.69838, 1.90430,
                 2.12928, 2.37557, 2.64563, 2.94217)
      x = (1.01, 1.02, 1.03, 1.11, 1.12, 1.13, 1.41, 1.42, 1.43, 1.44,
           1.45, 1.46, 1.75, 1.73, 1.77, 1.78, 1.79)
      display(Latex("Найти значения $sh(x)$ для значений аргумента:"))
      display(Latex("1) 1.01, 1.02, 1.03, 1.11, 1.12, 1.13"))
      display(Latex("2) 1.41, 1.42, 1.43, 1.44"))
      display(Latex("3) 1.45, 1.46, 1.75, 1.73, 1.77, 1.78, 1.79"))
      display(Latex("При этом дана таблица значений x_i \ u f_i=f(x_i):"))
      display(Latex("$i=0,\ldots,8;$"))
      \rightarrow 1.8\};$"))
      display(Latex("f_i)in{1.17520, , 1.33565, , 1.50946, , 1.69838, , 1.90430, , )
                 2.12928, \, 2.37557, \, 2.64563, \, 2.94217 \}; "))
      display(Latex("Ответы:"))
      section = 1
      for i, q in enumerate(x):
          if i == 0 or i == 6 or i == 10:
              display(Latex(f"{section}."))
              section += 1
          display(Latex(f"$P_{len}(x_nodes)) ({q}) = %.4f$"%
                       newton_polynomial(q, x_nodes, f_nodes)))
      l = np.linspace(1, 1.9)
      fig, ax = plt.subplots() # Create a figure and an axes.
      plt.style.use('seaborn-poster')
      ax.plot(1, sinh(1), label="\$sh(x)\$")# Plot some data on the axes.
      ax.set_xlabel('x label')
      ax.set_ylabel('y label')
```

```
ax.set_title("$y=sh(x)$")
ax.legend()
```

Найти значения sh(x) для значений аргумента:

- 1) 1.01, 1.02, 1.03, 1.11, 1.12, 1.13
- $2)\ 1.41,\ 1.42,\ 1.43,\ 1.44$
- 3) 1.45, 1.46, 1.75, 1.73, 1.77, 1.78, 1.79

При этом дана таблица значений x_i и $f_i = f(x_i)$:

$$i = 0, ..., 8;$$

 $x_i \in \{1.0, 1.1, 1.2, 1.3, 1.4, 1.5, 1.6, 1.7, 1.8\};$

 $f_i \in \{1.17520, 1.33565, 1.50946, 1.69838, 1.90430, 2.12928, 2.37557, 2.64563, 2.94217\};$

Ответы:

1.

- $P_9(1.01) = 1.1907$
- $P_9(1.02) = 1.2063$
- $P_9(1.03) = 1.2220$
- $P_9(1.11) = 1.3524$
- $P_9(1.12) = 1.3693$
- $P_9(1.13) = 1.3863$

2.

- $P_9(1.41) = 1.9259$
- $P_9(1.42) = 1.9477$
- $P_9(1.43) = 1.9697$
- $P_9(1.44) = 1.9919$

3.

- $P_9(1.45) = 2.0143$
- $P_9(1.46) = 2.0369$
- $P_9(1.75) = 2.7904$
- $P_9(1.73) = 2.7317$
- $P_9(1.77) = 2.8503$
- $P_9(1.78) = 2.8806$
- $P_9(1.79) = 2.9112$

[270]: <matplotlib.legend.Legend at 0x2add1510a60>

