

УНИВЕРСИТЕТ ИТМО

ВЫЧИСЛИТЕЛЬНАЯ МАТЕМАТИКА

Лабораторная работа №4

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Учебная группа: Р32101
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1 Цель работы

Найти функцию, являющуюся наилучшим приближением заданной табличной функции по методу наименьших квадратов.

2 Вычислительная реализация задачи

Функция по варианту:

$$y = \frac{4x}{x^4 + 3}, \quad x \in [-2; 0], \quad h = 0.2$$

2.1 Табулирование

x	-2.000	-1.800	-1.600	-1.400	-1.200	-1.000	-0.800	-0.600	-0.400	-0.200	0.000
y	-0.421	-0.533	-0.670	-0.819	-0.946	-1.000	-0.939	-0.767	-0.529	-0.267	0.000

2.2 Линейная аппроксимация

$$\varphi(x) = b + ax$$

$$SX = \sum_{i=1}^n x_i = -11.000$$

$$SXX = \sum_{i=1}^n x_i^2 = 15.400$$

$$SXY = \sum_{i=1}^n x_i \cdot y_i = 7.631$$

$$SY = \sum_{i=1}^n y_i = -6.890$$

$$\begin{cases} b \cdot SX + a \cdot SXX = SXY \\ bn + a \cdot SX = SY \end{cases} \Rightarrow \begin{cases} a = 0.178 \\ b = -0.468 \end{cases} \quad (1)$$

x	-2.000	-1.800	-1.600	-1.400	-1.200	-1.000	-0.800	-0.600	-0.400	-0.200	0.000
y	-0.421	-0.533	-0.670	-0.819	-0.946	-1.000	-0.939	-0.767	-0.529	-0.267	0.000
$\varphi(x)$	-0.795	-0.761	-0.727	-0.694	-0.660	-0.626	-0.593	-0.559	-0.525	-0.492	-0.458
ε	0.140	0.052	0.003	0.016	0.082	0.140	0.120	0.043	0.000	0.051	0.210

Среднеквадратичное отклонение: $\delta = 0.279$

2.3 Квадратичная аппроксимация

$$\varphi(x) = a_0 + a_1x + a_2x^2$$

$$SX = \sum_{i=1}^n x_i = -11.000$$

$$SXX = \sum_{i=1}^n x_i^2 = 15.400$$

$$SXXX = \sum_{i=1}^n x_i^3 = -24.200$$

$$SXXXX = \sum_{i=1}^n x_i^4 = 40.533$$

$$SXXY = \sum_{i=1}^n x_i^2 \cdot y_i = -10.066$$

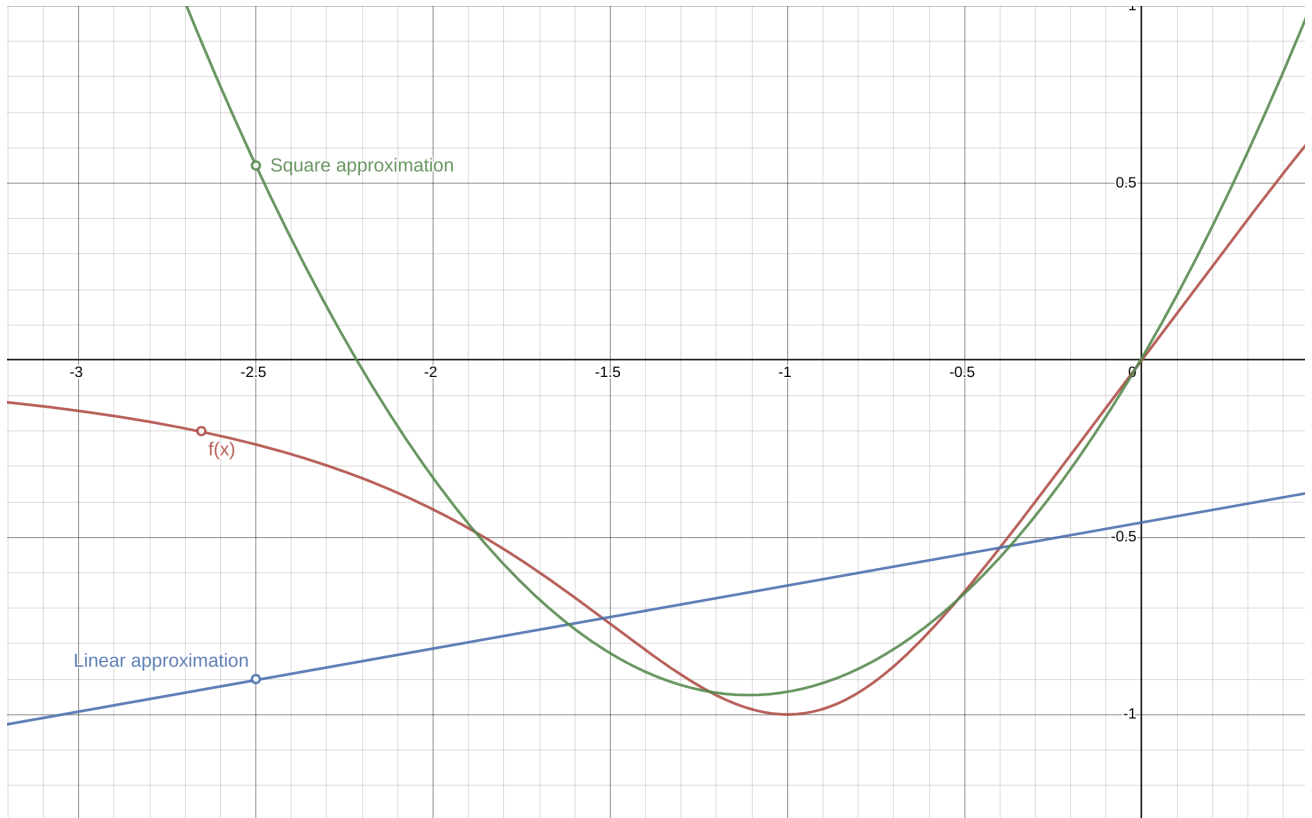
$$SXY = \sum_{i=1}^n x_i \cdot y_i = 7.631$$

$$SY = \sum_{i=1}^n y_i = -6.890$$

$$\begin{cases} a_0 n + a_1 \cdot SX + a_2 \cdot SXX = SY \\ a_0 \cdot SX + a_1 \cdot SXX + a_2 \cdot SXXX = SXY \\ a_0 \cdot SXX + a_1 \cdot SXXX + a_2 \cdot SXXXX = SXXY \end{cases} \Rightarrow \begin{cases} a_0 = 0.006 \\ a_1 = 1.715 \\ a_2 = 0.773 \end{cases} \quad (2)$$

Среднеквадратичное отклонение: $\delta = 0.055$

2.4 Графики



3 Программная реализация задачи

```
1 class ApproximationBuilder:
2     def __init__(self, x, y):
3         self.n = min(len(x), len(y))
4         self.x = x[:self.n]
5         self.y = y[:self.n]
6
7         self.sx = 0
8         self.sx2 = 0
9         self.sx3 = 0
10        self.sx4 = 0
11        self.sx5 = 0
12        self.sx6 = 0
13        self.sy = 0
14        self.syx = 0
15        self.syx2 = 0
16        self.syx3 = 0
17
18        for i in range(self.n):
19            self.sx += x[i]
20            self.sx2 += x[i] ** 2
21            self.sx3 += x[i] ** 3
22            self.sx4 += x[i] ** 4
23            self.sx5 += x[i] ** 5
24            self.sx6 += x[i] ** 6
25            self.sy += y[i]
26            self.syx += y[i] * x[i]
27            self.syx2 += y[i] * x[i] ** 2
28            self.syx3 += y[i] * x[i] ** 3
29
30        self.slnx = 0
31        self.slnx2 = 0
32        self.sylnx = 0
33        self.slny = 0
34        self.sxlny = 0
35        self.slnxlny = 0
36        try:
37            for i in range(self.n):
38                self.slnx += log(self.x[i])
39                self.slnx2 += log(self.x[i]) ** 2
40                self.sylnx += log(self.x[i]) * self.y[i]
41                self.slny += log(self.y[i])
42                self.sxlny += log(self.y[i]) * self.x[i]
43                self.slnxlny += log(self.y[i]) * log(self.x[i])
44        except ValueError:
45            self.slnx = None
46            self.slnx2 = None
47            self.sylnx = None
48            self.slny = None
49            self.sxlny = None
50            self.slnxlny = None
51
52        def linear(self):
53            matrix = Matrix(2,
54                            [
55                                [self.sx, self.sx2, self.syx],
56                                [self.n, self.sx, self.sy]
57                            ])
58            coeffs = gauss_solve(matrix).data
```

```

59     approx_y = []
60     for i in self.x:
61         approx_y.append(coeffs[0] + coeffs[1] * i)
62     return Results(False, "OK", {
63         "coeffs": coeffs, "approx_y": approx_y, "type": ApproximationType.LINEAR
64     })
65
66 def poly_square(self):
67     matrix = Matrix(3,
68         [
69             [self.n, self.sx, self.sx2, self.sy],
70             [self.sx, self.sx2, self.sx3, self.syx],
71             [self.sx2, self.sx3, self.sx4, self.syx2]
72         ])
73     coeffs = gauss_solve(matrix).data
74     approx_y = []
75     for i in self.x:
76         approx_y.append(coeffs[0] + coeffs[1] * i + coeffs[2] * i ** 2)
77     return Results(False, "OK", {
78         "coeffs": coeffs, "approx_y": approx_y, "type": ApproximationType.POLY_SQUARE
79     })
80
81 def poly_cubic(self):
82     matrix = Matrix(4,
83         [
84             [self.n, self.sx, self.sx2, self.sx3, self.sy],
85             [self.sx, self.sx2, self.sx3, self.sx4, self.syx],
86             [self.sx2, self.sx3, self.sx4, self.sx5, self.syx2],
87             [self.sx3, self.sx4, self.sx5, self.sx6, self.syx3]
88         ])
89     coeffs = gauss_solve(matrix).data
90     approx_y = []
91     for i in self.x:
92         approx_y.append(coeffs[0] + coeffs[1] * i + coeffs[2] * i ** 2 + coeffs[3] * i ** 3)
93     return Results(False, "OK", {
94         "coeffs": coeffs, "approx_y": approx_y, "type": ApproximationType.POLY_CUBIC
95     })
96
97 def exponential(self):
98     if None in [self.slny, self.sxlny]:
99         return Results(True,
100             "Cannot proceed exponential approximation for y < 0",
101             {"coeffs": [], "approx_y": []}
102         )
103     matrix = Matrix(2, [
104         [self.n, self.sx, self.slny],
105         [self.sx, self.sx2, self.sxlny]
106     ])
107     coeffs = gauss_solve(matrix).data
108     coeffs[0] = e ** coeffs[0]
109
110     approx_y = []
111     for i in self.x:
112         approx_y.append(coeffs[0] * e ** (coeffs[1] * i))
113     return Results(False, "OK", {
114         "coeffs": coeffs, "approx_y": approx_y, "type": ApproximationType.EXPONENTIAL
115     })
116
117 def logarithmic(self):
118     if None in [self.slnx, self.slnx2, self.sylnx]:
119         return Results(True,

```

```

120         "Cannot proceed logarithmic approximation for x < 0",
121         {"coeffs": [], "approx_y": []}
122     )
123
124     matrix = Matrix(2, [
125         [self.slnx2, self.slnx, self.sylnx],
126         [self.slnx, self.n, self.sy]
127     ])
128     coeffs = gauss_solve(matrix).data
129
130     approx_y = []
131     for i in self.x:
132         approx_y.append(coeffs[0] * log(i) + coeffs[1])
133     return Results(False, "OK", {
134         "coeffs": coeffs, "approx_y": approx_y, "type": ApproximationType.LOGARITHMIC
135     })
136
137 def power(self):
138     if None in [self.slnx, self.slnx2, self.slny, self.slnxlny]:
139         return Results(True,
140             "Cannot proceed power approximation for x < 0 or y < 0",
141             {"coeffs": [], "approx_y": []}
142         )
143
144     matrix = Matrix(2, [
145         [self.n, self.slnx, self.slny],
146         [self.slnx, self.slnx2, self.slnxlny]
147     ])
148     coeffs = gauss_solve(matrix).data
149     coeffs[0] = e ** coeffs[0]
150     approx_y = []
151     for i in self.x:
152         approx_y.append(coeffs[0] * i ** coeffs[1])
153     return Results(False, "OK", {
154         "coeffs": coeffs,
155         "approx_y": approx_y,
156         "type": ApproximationType.POWER
157     })
158
159

```

Полный код программы доступен по [ссылке](#).

4 Результат работы программы

```
1 [1] Keyboard
2 [2] File
3
4 Your choice: 2
5 Specify file name: 1
6 Cannot proceed exponential approximation for y < 0
7 Cannot proceed logarithmic approximation for x < 0
8 Cannot proceed power approximation for x < 0 or y < 0
9
10
11 Linear approximation
12 |
13 | Coefficients:
14 |
15 | \
16 |         -0.459, 0.168
17 |
18 | /
19 |
20 | Standard deviation:
21 |
22 | \
23 |         2.79e-01
24 |
25 | /
26 |
27 | Pearson correlation coefficient:
28 |
29 | \
30 |         3.57e-01
31 |
32 | /
33 |
34 |
35
36 -----
37 |      x      |      y      |      f(x)      |      eps      |
38 |-----|-----|-----|-----|
39 | -2.00e+00 | -4.21e-01 | -7.95e-01 | -3.74e-01 |
40 | -1.80e+00 | -5.33e-01 | -7.61e-01 | -2.28e-01 |
41 | -1.60e+00 | -6.70e-01 | -7.28e-01 | -5.78e-02 |
42 | -1.40e+00 | -8.19e-01 | -6.94e-01 | 1.25e-01 |
43 | -1.20e+00 | -9.46e-01 | -6.61e-01 | 2.85e-01 |
44 | -1.00e+00 | -1.00e+00 | -6.27e-01 | 3.73e-01 |
45 | -8.00e-01 | -9.39e-01 | -5.93e-01 | 3.46e-01 |
46 | -6.00e-01 | -7.67e-01 | -5.60e-01 | 2.07e-01 |
47 | -4.00e-01 | -5.29e-01 | -5.26e-01 | 2.80e-03 |
48 | -2.00e-01 | -2.67e-01 | -4.93e-01 | -2.26e-01 |
49 | 0.00e+00 | 0.00e+00 | -4.59e-01 | -4.59e-01 |
50 |-----|-----|-----|-----|
51
52
53
54 Square polynomial approximation
55 |
56 | Coefficients:
57 |
58 | \
```

```

59 |          0.006, 1.716, 0.774
60 |
61 | /
62 |
63 | Standard deviation:
64 |
65 | \
66 |          5.48e-02
67 |
68 | /
69 |
70 |

```

x	y	f(x)	eps
-2.00e+00	-4.21e-01	-3.30e-01	9.10e-02
-1.80e+00	-5.33e-01	-5.75e-01	-4.20e-02
-1.60e+00	-6.70e-01	-7.58e-01	-8.82e-02
-1.40e+00	-8.19e-01	-8.79e-01	-6.04e-02
-1.20e+00	-9.46e-01	-9.39e-01	7.36e-03
-1.00e+00	-1.00e+00	-9.36e-01	6.40e-02
-8.00e-01	-9.39e-01	-8.71e-01	6.76e-02
-6.00e-01	-7.67e-01	-7.45e-01	2.20e-02
-4.00e-01	-5.29e-01	-5.57e-01	-2.76e-02
-2.00e-01	-2.67e-01	-3.06e-01	-3.92e-02
0.00e+00	0.00e+00	6.00e-03	6.00e-03

```

87
88
89
90 Cubic polynomial approximation

```

```

91 |
92 | Coefficients:
93 |
94 | \
95 |          0.053, 2.085, 1.257, 0.161
96 |
97 | /
98 |
99 | Standard deviation:
100 |
101 | \
102 |          4.55e-02
103 |
104 | /
105 |
106 |

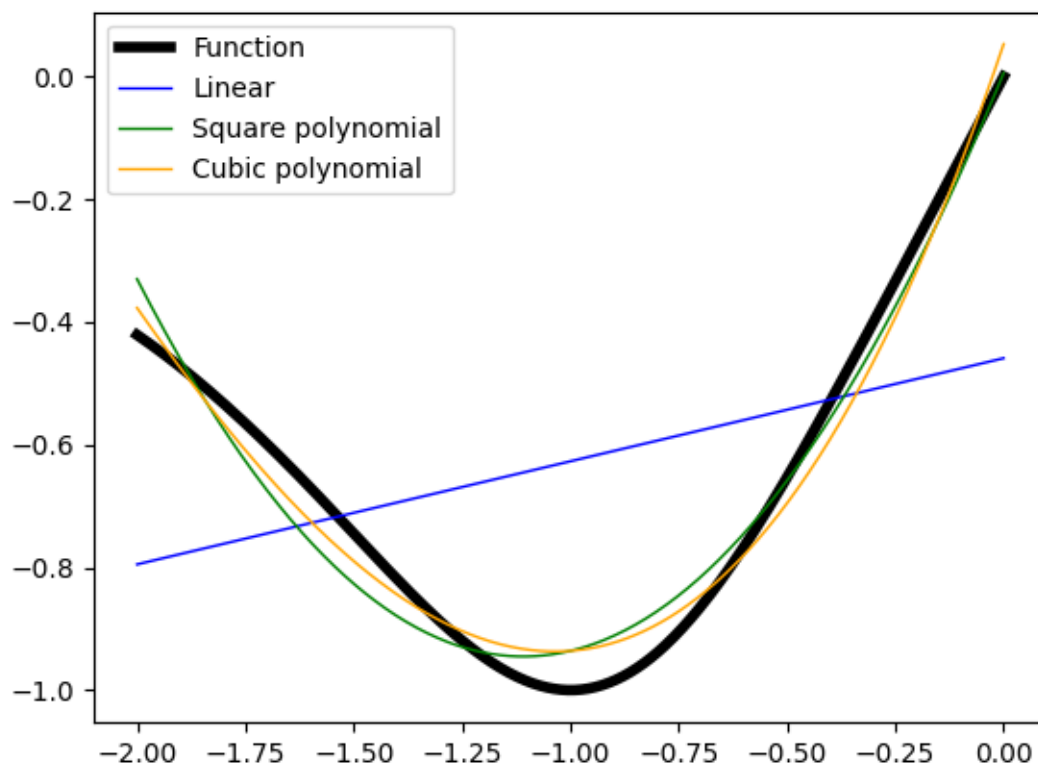
```

x	y	f(x)	eps
-2.00e+00	-4.21e-01	-3.77e-01	4.40e-02
-1.80e+00	-5.33e-01	-5.66e-01	-3.33e-02
-1.60e+00	-6.70e-01	-7.25e-01	-5.45e-02
-1.40e+00	-8.19e-01	-8.44e-01	-2.51e-02
-1.20e+00	-9.46e-01	-9.17e-01	2.89e-02
-1.00e+00	-1.00e+00	-9.36e-01	6.40e-02
-8.00e-01	-9.39e-01	-8.93e-01	4.60e-02
-6.00e-01	-7.67e-01	-7.80e-01	-1.33e-02
-4.00e-01	-5.29e-01	-5.90e-01	-6.12e-02


```

120 | -2.00e-01 | -2.67e-01 | -3.15e-01 | -4.80e-02 |
121 | 0.00e+00 | 0.00e+00 | 5.30e-02 | 5.30e-02 |
122 | ----- | ----- | ----- | ----- |
123
124 -----
125 Best approximation: Cubic polynomial
126 -----
127

```



5 Вывод

Для наиболее оптимального распределения человеческих ресурсов и поддержания количества нервных клеток в головном мозгу, настоятельно рекомендуется использовать готовые библиотеки, содержащие наиболее эффективные реализации алгоритмов, вместо самостоятельной реализации оных.

Иными словами, вообще лучше зачилиться и не изобретать велосипед.