

Университет ИТМО
МФ КТиУ, Ф ПИиКТ

Лабораторная работа №4
Дисциплина «Вычислительная математика»

Аппроксимация функции методом наименьших квадратов

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Вычислительная реализация задачи

Линейная аппроксимация:

$$y = \frac{12x}{x^4 + 1}$$

$$n = 11$$

$$x \text{ in } [0; 2]$$

$$h = 0.2$$

i	1	2	3	4	5	6	7	8	9	10	11
x_i	-2	-1.8	-1.6	-1.4	-1.2	-1	-0.8	-0.6	-0.4	-0.2	0
y_i	-1.412	1.879	-2.542	-3.47	-4.685	-6.0	-6.81	-6.374	-4.68	-2.396	0.0

$$\varphi(x) = a + b * x$$

Вычисляем суммы: $sx = -11$, $sxx = 15.4$, $sy = -40.248$, $sxy = 38.377$

$$\begin{cases} n * a + sx * b = sy \\ sx * a + sxx * b = sxy \end{cases} \begin{cases} 11 * a - 11 * b = -40.248 \\ -11 * a + 15.4 * b = 38.377 \end{cases}$$

$$\begin{cases} a = -4.084 \\ b = -0.425 \end{cases}$$

$$\varphi(x) = -4.084 - 0.425 * x$$

$$x_i \quad -2.0 \quad -1.8 \quad -1.6 \quad -1.4 \quad -1.2 \quad -1.0 \quad -0.8 \quad -0.6 \quad -0.4 \quad -0.2 \quad 0.0$$

$$y_i \quad -1.412 \quad -1.879 \quad -2.542 \quad -3.47 \quad -4.685 \quad -6.0 \quad -6.81 \quad -6.374 \quad -4.68 \quad -2.396 \quad 0.0$$

$$\varphi(x_i) \quad 7.743 \quad 6.926 \quad 6.109 \quad 5.293 \quad 4.476 \quad 3.659 \quad 2.842 \quad 2.025 \quad 1.209 \quad 0.392 \quad -0.425$$

$$\text{delta} \quad 83.814 \quad 77.528 \quad 74.84 \quad 76.79 \quad 83.924 \quad 93.296 \quad 93.161 \quad 70.543 \quad 34.68 \quad 7.773 \quad 0.181$$

$$\sigma = \sqrt{\frac{\sum (\varphi(x_i) - y_i)^2}{n}} = 2.602$$

Квадратичная аппроксимация:

$$y = \frac{12x}{x^4 + 1}$$

$$n = 11$$

$$x \text{ in } [0; 2]$$

$$h = 0.2$$

i	1	2	3	4	5	6	7	8	9	10	11
x_i	-2	-1.8	-1.6	-1.4	-1.2	-1	-0.8	-0.6	-0.4	-0.2	0
y_i	-1.412	1.879	-2.542	-3.47	-4.685	-6.0	-6.81	-6.374	-4.68	-2.396	0.0

$$\varphi(x) = a + b * x + c * x^2$$

Вычисляем суммы: $sx = -11$, $sxx = 15.4$, $sxxx = -24.2$, $sxxxx = 40.533$,

$sy = -40.248$, $sxy = 38.377$, $sxxy = -45.289$

$$\begin{cases} n * a + sx * b + sxx * c = sy \\ sx * a + sxx * b + sxxx * c = sxy \\ sxx * a + sxxx * b + sxxxx * c = sxxxy \end{cases} \begin{cases} 11 * a - 11 * b + 15.4 * c = -40.248 \\ -11 * a + 15.4 * b - 24.2 * c = 38.377 \\ 15.4 * -24.2 * b + 40.533 * c = -45.289 \end{cases}$$

$$\begin{cases} 11 * a = -6.9 + 11 * b - 15.4 * c \\ 4.4 * b - 8.8 * c = 0.73 \\ 15.4 * a - 24.2 * b + 40.53 * c = -10.06 \end{cases} \begin{cases} a = -0.63 + b - 1.4 * c \\ 4.4 * b - 8.8 * c = 0.73 \\ 15.4 * a - 24.2 * b + 40.53 * c = -10.06 \end{cases}$$

$$\begin{cases} a = -0.63 + b - 1.4 * c \\ 4.4 * b - 8.8 * c = 0.73 \\ -9.7 + 15.4 * b - 21.56 * c - 24.2 * b + 40.53 * c = -10.06 \end{cases}$$

$$\begin{cases} a = -0.63 + b - 1.4 * c \\ 4.4 * b - 8.8 * c = 0.73 \\ -8.8 * b + 18.97 * c = -0.36 \end{cases} \begin{cases} a = -0.63 + b - 1.4 * c \\ 4.4 * b = 0.73 + 8.8 * c \\ -1.46 - 17.6 * c + 18.97 * c = -0.36 \end{cases}$$

$$\begin{cases} a = -0.63 + b - 1.4 * c \\ 4.4 * b = 0.73 + 8.8 * c \\ 1.37 * c = 1.1 \end{cases} \begin{cases} a = -0.63 + b - 1.4 * c \\ 4.4 * b = 7.77 \\ 1.37 * c = 1.1 \end{cases} \begin{cases} a = -0.63 + 1.77 - 1.4 * 0.8 \\ b = 1.77 \\ 1.37 * c = 1.1 \end{cases}$$

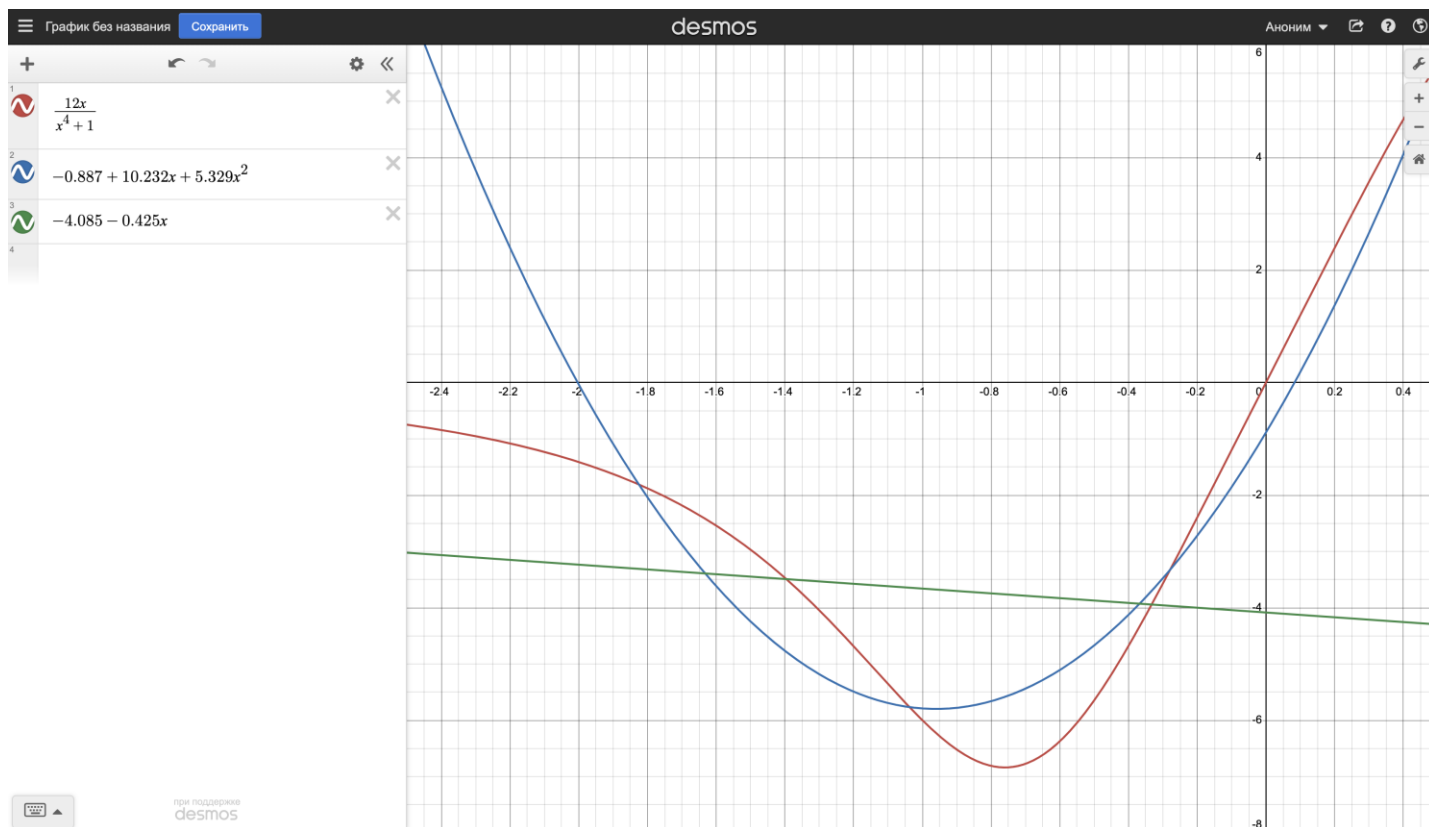
$$\begin{cases} a = -0.887 \\ b = 10.232 \\ c = 5.329 \end{cases}$$

$$\varphi(x) = 5.329 + 10.232 * x - 0.887 * x^2$$

x_i	-2.0	-1.8	-1.6	-1.4	-1.2	-1.0	-0.8	-0.6	-0.4	-0.2	0.0
y_i	-1.412	-1.879	-2.542	-3.47	-4.685	-6.0	-6.81	-6.374	-4.68	-2.396	0.0
ϕ_i	-0.035	-2.039	-3.616	-4.767	-5.492	-5.79	-5.662	-5.108	-4.127	-2.72	-0.887
δ_i	1.896	0.026	1.153	1.682	0.651	0.044	1.318	1.603	0.306	0.105	0.787

$$\sigma = \sqrt{\frac{\sum(\varphi(x_i) - y_i)^2}{n}} = 0.933$$

У квадратичной аппроксимации среднеквадратичное отклонение меньше, поэтому это приближение наилучшее.



Программная реализация задачи

```
import inspect
from math import sqrt, exp, log

import matplotlib.pyplot as plt

def add_col(m, col):
    for k, row in enumerate(m):
        row.append(col[k])
    return m

def remove_last_col(m):
    for k, row in enumerate(m):
        row.pop()
    return m

def plus(src, ind, m):
    for i in range(src + 1, len(m)):
        _plus(src, i, m, -m[i][ind] / m[src][ind])

def _plus(src, dest, m, mul: float = 1):
    for i in range(len(m[0])):
        m[dest][i] += m[src][i] * mul

def swap(src, dest, m):
    m[src], m[dest] = m[dest], m[src]

def rang(m):
    return sum(any(row) for row in m)
```

```

def determinant(m, k):
    p = 1
    for i in range(len(m)):
        p *= m[i][i]
    return (-1) ** k * p

def solve(m):
    k = 0
    row = 0
    col = 0
    n = len(m)
    while col < n:
        for j in range(row, n):
            if m[j][col]:
                swap(j, row, m)
                k += row != j
                plus(row, col, m)
                row += 1
                break
        col += 1
    xs = []
    for i in range(len(m)):
        x = m[len(m) - i - 1][-1]
        for j in range(1, i + 2):
            if j == i + 1:
                x /= m[len(m) - i - 1][-j - 1]
            else:
                x -= m[len(m) - i - 1][-j - 1] * xs[j - 1]
        xs.append(x)
    return xs[::-1]

def linear_approximation(xs, ys, n):
    sx = sum(xs)
    sxx = sum(x ** 2 for x in xs)
    sy = sum(ys)
    sxy = sum(x * y for x, y in zip(xs, ys))
    ext_matrix = add_col(
        [
            [n, sx],
            [sx, sxx]
        ],
        [sy, sxy]
    )

    a, b = solve(ext_matrix)
    return lambda x: a + b * x, a, b

def quadratic_approximation(xs, ys, n):
    sx = sum(xs)
    sxx = sum(x ** 2 for x in xs)
    sxxx = sum(x ** 3 for x in xs)
    sxxxx = sum(x ** 4 for x in xs)
    sy = sum(ys)
    sxy = sum(x * y for x, y in zip(xs, ys))
    sxxxy = sum(x * x * y for x, y in zip(xs, ys))
    ext_matrix = add_col(
        [
            [n, sx, sxx],
            [sx, sxx, sxxx],
            [sxx, sxxx, sxxxx]
        ],
        [sy, sxy, sxxxy]
    )

    a, b, c = solve(ext_matrix)

```

```

    return lambda x: a + b * x + c * x ** 2, a, b, c

def cubic_approximation(xs, ys, n):
    sx = sum(xs)
    sxx = sum(x ** 2 for x in xs)
    sxxx = sum(x ** 3 for x in xs)
    sxxxx = sum(x ** 4 for x in xs)
    sxxxxx = sum(x ** 5 for x in xs)
    sxxxxxx = sum(x ** 6 for x in xs)
    sy = sum(ys)
    sxy = sum(x * y for x, y in zip(xs, ys))
    sxxxy = sum(x * x * y for x, y in zip(xs, ys))
    sxxxxy = sum(x * x * x * y for x, y in zip(xs, ys))
    ext_matrix = add_col(
        [
            [n, sx, sxx, sxxx],
            [sx, sxx, sxxx, sxxxx],
            [sxx, sxxx, sxxxx, sxxxxx],
            [sxxx, sxxxx, sxxxxx, sxxxxxx]
        ],
        [sy, sxy, sxxxy, sxxxxy]
    )

    a, b, c, d = solve(ext_matrix)
    return lambda x: a + b * x + c * x ** 2 + d * x ** 3, \
        a, b, c, d

def exponential_approximation(xs, ys, n):
    ys_ = list(map(log, ys))
    _, a_, b_ = linear_approximation(xs, ys_, n)
    a = exp(a_)
    b = b_
    return lambda x: a * exp(b * x), a, b

def logarithmic_approximation(xs, ys, n):
    xs_ = list(map(log, xs))
    _, a_, b_ = linear_approximation(xs_, ys, n)
    a = a_
    b = b_
    return lambda x: a + b * log(x), a, b

def power_approximation(xs, ys, n):
    xs_ = list(map(log, xs))
    ys_ = list(map(log, ys))
    _, a_, b_ = linear_approximation(xs_, ys_, n)
    a = exp(a_)
    b = b_
    return lambda x: a * x ** b, a, b

def calc_deviation(xs, ys, fi, n):
    return sum((eps ** 2 for eps in [fi(x) - y for x, y in zip(xs, ys)]))

def calc_standard_deviation(xs, ys, fi, n):
    return sqrt(sum(((fi(x) - y) ** 2 for x, y in zip(xs, ys))) / n)

def calc_pearson_correlation_coefficient(xs, ys, n):
    av_x = sum(xs) / n
    av_y = sum(ys) / n
    return sum((x - av_x) * (y - av_y) for x, y in zip(xs, ys)) / \
        sqrt(sum((x - av_x) ** 2 for x in xs) *
              sum((y - av_y) ** 2 for y in ys))

```

```

def calc_coefficient_of_determination(xs, ys, fi, n):
    return 1 - sum((y - fi(x)) ** 2 for x, y in zip(xs, ys)) / (sum(fi(x) ** 2 for x in
xs) - sum(fi(x) for x in xs) ** 2 / n)

def get_str_content_of_func(func):
    str_func = inspect.getsourcelines(func)[0][0]
    return str_func.split('lambda x: ')[-1].split(',')[0].strip()

def draw_plot(a, b, func, dx=0.1):
    xs, ys = [], []
    a -= dx
    b += dx
    x = a
    while x <= b:
        xs.append(x)
        ys.append(func(x))
        x += dx
    plt.plot(xs, ys, 'g')

def read_number(s: str):
    while True:
        try:
            return float(input(s))
        except Exception:
            continue

if __name__ == '__main__':
    read_number("Введите количество точек: ")
    xs = list(map(float, input().split()))
    ys = list(map(float, input().split()))
    if len(xs) != len(ys):
        print("Некорректные данные")

    n = len(xs)

    names = {
        linear_approximation: "Линейная",
        power_approximation: "Степенная",
        exponential_approximation: "Экспоненциальная",
        logarithmic_approximation: "Логарифмическая",
        quadratic_approximation: "Квадратичная",
        cubic_approximation: "Кубическая"
    }

    if all(map(lambda x: x > 0, xs)) and all(map(lambda x: x > 0, ys)):
        approximation_funcs = [
            linear_approximation,
            power_approximation,
            exponential_approximation,
            logarithmic_approximation,
            quadratic_approximation,
            cubic_approximation
        ]
    else:
        approximation_funcs = [
            linear_approximation,
            quadratic_approximation,
            cubic_approximation
        ]

    best_sigma = float('inf')
    best_aprxmt_f = None

    for aprxmt_f in approximation_funcs:
        print(names[aprxmt_f], ": ")
        fi, *coeffs = aprxmt_f(xs, ys, n)

```

```

s = calc_deviation(xs, ys, fi, n)
sigma = calc_standard_deviation(xs, ys, fi, n)
if sigma < best_sigma:
    best_sigma = sigma
    best_aprxmt_f = aprxmt_f
r2 = calc_coefficient_of_determination(xs, ys, fi, n)
print('fi(x) =', get_str_content_of_func(fi))
print(f'coeffs:', list(map(lambda cf: round(cf, 4), coeffs)))
print(f'S = {s:.5f}, σ = {sigma:.5f}, R2 = {r2:.5f}')
if aprxmt_f is linear_approximation:
    print('r =', calc_pearson_correlation_coefficient(xs, ys, n))
plt.title(names[aprxmt_f])

draw_plot(xs[0], xs[-1], fi)
for i in range(n):
    plt.scatter(xs[i], ys[i], c='r')
plt.xlabel("X")
plt.ylabel("Y")
plt.show()
print('-' * 50)
print(f'best_func: {names[best_aprxmt_f]}')

```

Тестовые данные

→ lab4 git:(main) × python3 main.py

Введите количество точек: 11

-2.0	-1.8	-1.6	-1.4	-1.2	-1.0	-0.8	-0.6	-0.4	-0.2	0.0
-1.412	-1.879	-2.542	-3.47	-4.685	-6.0	-6.81	-6.374	-4.68	-2.396	0.0

Линейная :

$fi(x) = a + b * x$

coeffs: [-4.0841, -0.4252]

S = 48.56409, σ = 2.10117, R2 = -60.04082

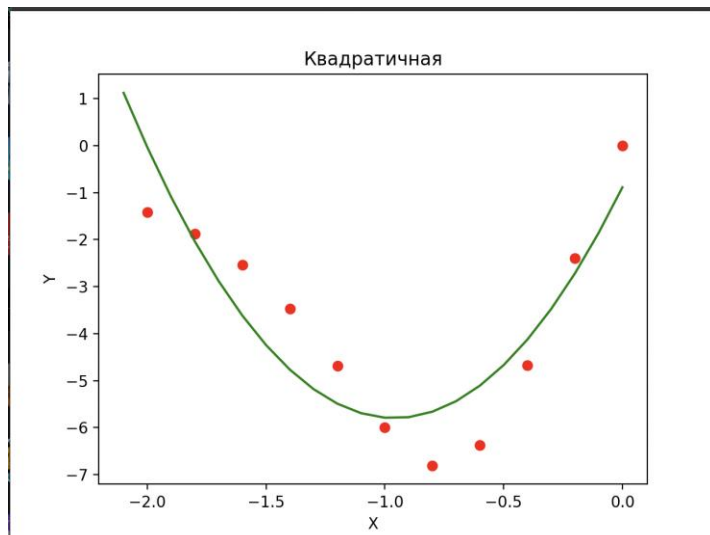
r = -0.12695833690235026

Квадратичная :

$fi(x) = a + b * x + c * x ** 2$

coeffs: [-0.8864, 10.2339, 5.3296]

S = 9.57058, σ = 0.93277, R2 = 0.75947



Линейная :

$$f_l(x) = a + b * x$$

coeffs: [-4.0841, -0.4252]

S = 48.56409, σ = 2.10117, R2 = -60.04082

r = -0.12695833690235026

Квадратичная :

$$f_l(x) = a + b * x + c * x ** 2$$

coeffs: [-0.8864, 10.2339, 5.3296]

S = 9.57058, σ = 0.93277, R2 = 0.75947

Кубическая :

$$f_l(x) = a + b * x + c * x ** 2 + d * x ** 3$$

coeffs: [0.4356, 20.7361, 19.1, 4.5901]

S = 1.24048, σ = 0.33581, R2 = 0.97422

best_func: Кубическая

Вывод

В ходе лабораторной работы я познакомился с аппроксимацией функции методом наименьших квадратов.

Метод наименьших квадратов – хороший метод, определяются параметры, при которых значения аппроксимирующей функции приблизительно совпадали бы со значениями исходной функции. В качестве аппроксимирующих функций обычно берут многочлены. Чем выше степень многочлена, тем точнее. Можно использовать экспоненциальные, логарифмические, степенные функции (сводя их преобразованиями к линейному аппроксимированию). Аппроксимирующая функций проходит в ближайшей близости от точек из заданного массива данных.