

Aproksymacja metoda najmniejszych kwadratów

x_i	$f(x) = y$
x_1	$f(x_1) = y_1$
x_2	$f(x_2) = y_2$
\vdots	
x_n	$f(x_n) = y_n$

$F(x; \alpha_0, \dots, \alpha_m)$ - funkcja aproksymująca

$$S(\alpha_0, \dots, \alpha_m) = \sum_{i=1}^n (F(x_i; \alpha_0, \dots, \alpha_m) - y_i)^2$$



ta wartość jak najmniejsza

$$\frac{\partial S}{\partial \alpha_j} = 0$$

min.

$$j = 0, \dots, m$$

$$\frac{\partial^2 S}{\partial \alpha_j^2} > 0$$

1. Metoda liniowa: F jest liniowe
względem $\alpha_0, \dots, \alpha_m$ czyli

$$F(X; \alpha_0, \dots, \alpha_m) = \sum_{k=0}^m \alpha_k X^k =$$

$$= \alpha_0 + \alpha_1 X + \alpha_2 X^2 + \dots + \alpha_m X^m$$

2. Metoda nieliniowa: F jest nieliniowe
względem $\alpha_0, \dots, \alpha_m$, np.

$$F(X; \alpha_0, \alpha_1, \alpha_2) = \frac{\alpha_0}{1 + 4(X - \alpha_1)^2 / \alpha_2^2}$$

$$S(\alpha_0 \dots \alpha_m) = \sum_{i=1}^n \left(\sum_{k=0}^m \alpha_k X_i^k - y_i \right)^2$$

$$\frac{\partial S}{\partial \alpha_j} = 0, \quad j = 0, 1, \dots, m$$

$$\begin{aligned} \frac{\partial S}{\partial \alpha_j} &= \frac{\partial}{\partial \alpha_j} \sum_{i=1}^n \left(\sum_{k=0}^m \alpha_k X_i^k - y_i \right)^2 = \\ &= \sum_{i=1}^n 2 \left(\sum_{k=0}^m \alpha_k X_i^k - y_i \right) X_i^j, \quad \text{so} \end{aligned}$$

$$\begin{aligned} \frac{\partial}{\partial \alpha_j} \sum_{k=0}^m \alpha_k X_i^k &= \sum_{k=0}^m \frac{\partial \alpha_k}{\partial \alpha_j} X_i^k = \\ &= \sum_{k=0}^m \delta_{kj} X_i^k = X_i^j. \end{aligned}$$

Crqbi

$$\sum_{i=1}^n \left(\sum_{k=0}^m \alpha_k X_i^{k+\delta_j} - X_i^{\delta_j} y_i \right) = 0$$

$j = 0, 1, \dots, m$

$$j = 0, 1, \dots, m$$

$$\sum_{i=1}^n \sum_{k=0}^m \alpha_k X_i^{k+j} = \sum_{i=1}^n X_i^j y_i$$

lub

$$\sum_{k=0}^m \alpha_k \sum_{i=1}^n X_i^{k+j} = \sum_{i=1}^n X_i^j y_i$$

what $m+1$ vectors X_i^{k+j}
na $m+1$ niewiadomych α_j

Ad. 2

$$S(\alpha_0, \dots, \alpha_m) = \sum_{i=1}^n (F(x_i; \alpha_0, \dots, \alpha_m) - y_i)^2$$

Nasze zadanie sprowadza się
do znalezienia minimum funkcji S .
Np. pierwszy sposób

$\alpha_0^0, \alpha_1^0, \dots, \alpha_m^0$ — zerowe przybliżenie

↓

$\alpha_0^1, \alpha_1^1, \dots, \alpha_m^1$

$\alpha_0^2, \alpha_1^2, \dots, \alpha_m^2$

⋮

$\alpha_0^k, \alpha_1^k, \dots, \alpha_m^k$ — pierwsze przybliżenie

↓

Ad

14. Drugi sposób prowadzi do ułamku r -i' binomialy A6

$$O_{2m} \quad \alpha_0, \alpha_1, \dots, \alpha_m = \alpha \quad ; \quad \alpha_0, \alpha_1, \dots, \alpha_m = \alpha$$

$$x \frac{f_{pe}}{(x)se} - \frac{f_{pe}}{(x)se} + \frac{f_{pe}}{(x)se} = 0$$

$$\frac{c_{pe}}{c_{se}} \rho_m \frac{d}{dt} (\rho_m - \rho) + \dots + \frac{c_{pe}}{c_{se}} v_{pe} \frac{d}{dt} (v_{pe} - v)$$

[illegible]

$$c_{45} = \frac{c_{20e}}{(0.8)5e}$$

$$S_{j0} + S_{j1} + \dots + S_{jm} = S_j \quad j=0,1,\dots,m$$

Wład i + 1 rożniam
na 5000, 500

$$\underline{f'(x) = \frac{df(x)}{dx} = \frac{f(x+h) - f(x-h)}{2h}}$$

$$\underline{S_0(\alpha_0^0, \alpha_1^0, \alpha_2^0) = \frac{\partial S(\alpha_0^0, \alpha_1^0, \alpha_2^0)}{\partial \alpha_0} =}$$

$$= \frac{S(\alpha_0^0 + \delta_0, \alpha_1^0, \alpha_2^0) - S(\alpha_0^0 - \delta_0, \alpha_1^0, \alpha_2^0)}{2\delta_0}$$

$$\underline{S_1(\alpha_0^0, \alpha_1^0, \alpha_2^0) = \frac{\partial S(\alpha_0^0, \alpha_1^0, \alpha_2^0)}{\partial \alpha_1} =}$$

$$= \frac{S(\alpha_0^0, \alpha_1^0 + \delta_1, \alpha_2^0) - S(\alpha_0^0, \alpha_1^0 - \delta_1, \alpha_2^0)}{2\delta_1}$$

$$\underline{S_2(\alpha_0^0, \alpha_1^0, \alpha_2^0) = \frac{\partial S(\alpha_0^0, \alpha_1^0, \alpha_2^0)}{\partial \alpha_2} =}$$

$$= \frac{S(\alpha_0^0, \alpha_1^0, \alpha_2^0 + \delta_2) - S(\alpha_0^0, \alpha_1^0, \alpha_2^0 - \delta_2)}{2\delta_2}$$

Elementary diagonal

$$f''(x) = \frac{d^2 f(x)}{dx^2} = \frac{f(x+h) - 2f(x) + f(x-h)}{h^2}$$

$$\alpha^0 = \alpha_0^0, \alpha_1^0, \alpha_2^0$$

$$\text{np. } S_{00}(\alpha_0^0, \alpha_1^0, \alpha_2^0) = \frac{\partial^2 S(\alpha^0)}{\partial \alpha_0^2} =$$

$$= \frac{S(\alpha_0^0 + \delta_0, \alpha_1^0, \alpha_2^0) - 2S(\alpha_0^0, \alpha_1^0, \alpha_2^0) + S(\alpha_0^0 - \delta_0, \alpha_1^0, \alpha_2^0)}{\delta_0^2}$$

$$S_0(x^0) = \frac{\partial}{\partial x^0} \frac{\partial \mathcal{L}}{\partial x^0} = \frac{\partial}{\partial x^0} \left(\frac{\partial \mathcal{L}}{\partial x^0} \right) = \frac{\partial}{\partial x^0} \left(\frac{\partial \mathcal{L}}{\partial x^0} \right) = \frac{\partial}{\partial x^0} \left(\frac{\partial \mathcal{L}}{\partial x^0} \right)$$

$$= \frac{1}{2\sigma_1} \left(\frac{\partial \mathcal{L}}{\partial x^0} + \frac{\partial \mathcal{L}}{\partial x^1} + \frac{\partial \mathcal{L}}{\partial x^2} \right) = \frac{\partial \mathcal{L}}{\partial x^0} + \frac{\partial \mathcal{L}}{\partial x^1} + \frac{\partial \mathcal{L}}{\partial x^2}$$

$$= \frac{\partial \mathcal{L}}{\partial x^0} + \frac{\partial \mathcal{L}}{\partial x^1} + \frac{\partial \mathcal{L}}{\partial x^2} = \frac{\partial \mathcal{L}}{\partial x^0} + \frac{\partial \mathcal{L}}{\partial x^1} + \frac{\partial \mathcal{L}}{\partial x^2}$$

Oznaczenia dla $i, j = 0, 1, 2$

$$\frac{\partial S}{\partial \alpha_i} = S_i$$

$$\frac{\partial}{\partial \alpha_k} \frac{\partial S}{\partial \alpha_j} = \frac{\partial^2 S}{\partial \alpha_k \partial \alpha_j} = \underbrace{S_{kj} = S_{jk}}$$

$$\alpha_i - \alpha_i^0 = h_i$$

$$(1) h_0 S_{00} + h_1 S_{01} + h_2 S_{02} = -S_0$$

$$(2) h_0 S_{10} + h_1 S_{11} + h_2 S_{12} = -S_1$$

$$(3) h_0 S_{20} + h_1 S_{21} + h_2 S_{22} = -S_2$$

Stąd h_0, h_1, h_2 czyli $\alpha_0^{(1)}, \alpha_1^{(1)}, \alpha_2^{(1)}$

Obliczać nowe S_{ij} i rozwiązać układ r-ú otrzymując lepsze $\alpha_0^{(2)}, \alpha_1^{(2)}, \alpha_2^{(2)}$ itd, aż do uzyskania określonej dokładności

Przykład dla $F(x_0, x_1, x_2)$

$$S(x_0, x_1, x_2) = \sum_{i=1}^n (F(x_i; x_0, x_1, x_2) - y_i)^2$$

Z warunku koniecznego istnieją minimum

$$\frac{\partial S(x_0, x_1, x_2)}{\partial x_0} = 0, \quad \frac{\partial S(x_0, x_1, x_2)}{\partial x_1} = 0, \quad \frac{\partial S(x_0, x_1, x_2)}{\partial x_2} = 0$$

Z warunku koniecznego Taylora wokół x_0^0, x_1^0, x_2^0

$$(V) \quad \frac{\partial S(x_0, x_1, x_2)}{\partial x_0} \approx \frac{\partial S(x_0^0, x_1^0, x_2^0)}{\partial x_0} + \frac{\partial^2 S(x_0^0, x_1^0, x_2^0)}{\partial x_0^2} (x_0 - x_0^0) + \frac{\partial^2 S(x_0^0, x_1^0, x_2^0)}{\partial x_0 \partial x_1} (x_1 - x_1^0) + \frac{\partial^2 S(x_0^0, x_1^0, x_2^0)}{\partial x_0 \partial x_2} (x_2 - x_2^0) + \dots$$

$$+ (x_1 - x_1^0) \frac{\partial^2 S(x_0^0, x_1^0, x_2^0)}{\partial x_1^2} + (x_2 - x_2^0) \frac{\partial^2 S(x_0^0, x_1^0, x_2^0)}{\partial x_2^2} + \dots$$

PRZYKŁAD 1.2.2. Regresja kwadratowa

$$F(x; \alpha_0, \alpha_1, \alpha_2) = \alpha_0 + \alpha_1 x + \alpha_2 x^2$$

$$S(\alpha_0, \alpha_1, \alpha_2) = \sum_{i=1}^n (\alpha_0 + \alpha_1 x_i + \alpha_2 x_i^2 - y_i)^2, \quad \frac{\partial S}{\partial \alpha_0} = \frac{\partial S}{\partial \alpha_1} = \frac{\partial S}{\partial \alpha_2} = 0 \quad (*)$$

$$\frac{\partial S}{\partial \alpha_0} = \sum_{i=1}^n 2(\alpha_0 + \alpha_1 x_i + \alpha_2 x_i^2 - y_i) \cdot \frac{\partial}{\partial \alpha_0} (\alpha_0 + \alpha_1 x_i + \alpha_2 x_i^2 - y_i) = 2n\alpha_0 + 2\left(\sum_{i=1}^n x_i\right)\alpha_1 + 2\left(\sum_{i=1}^n x_i^2\right)\alpha_2 - 2\sum_{i=1}^n y_i$$

$$\frac{\partial S}{\partial \alpha_1} = \sum_{i=1}^n 2(\alpha_0 + \alpha_1 x_i + \alpha_2 x_i^2 - y_i) \cdot \frac{\partial}{\partial \alpha_1} (\alpha_0 + \alpha_1 x_i + \alpha_2 x_i^2 - y_i) = 2\left(\sum_{i=1}^n x_i\right)\alpha_0 + 2\left(\sum_{i=1}^n x_i^2\right)\alpha_1 + 2\left(\sum_{i=1}^n x_i^3\right)\alpha_2 - 2\sum_{i=1}^n x_i y_i$$

$$\frac{\partial S}{\partial \alpha_2} = \sum_{i=1}^n 2(\alpha_0 + \alpha_1 x_i + \alpha_2 x_i^2 - y_i) \cdot \frac{\partial}{\partial \alpha_2} (\alpha_0 + \alpha_1 x_i + \alpha_2 x_i^2 - y_i) = 2\left(\sum_{i=1}^n x_i^2\right)\alpha_0 + 2\left(\sum_{i=1}^n x_i^3\right)\alpha_1 + 2\left(\sum_{i=1}^n x_i^4\right)\alpha_2 - 2\sum_{i=1}^n x_i^2 y_i$$

$$2(*) \quad \begin{cases} n\alpha_0 + \left(\sum_{i=1}^n x_i\right)\alpha_1 + \left(\sum_{i=1}^n x_i^2\right)\alpha_2 = \sum_{i=1}^n y_i \\ \left(\sum_{i=1}^n x_i\right)\alpha_0 + \left(\sum_{i=1}^n x_i^2\right)\alpha_1 + \left(\sum_{i=1}^n x_i^3\right)\alpha_2 = \sum_{i=1}^n x_i y_i \\ \left(\sum_{i=1}^n x_i^2\right)\alpha_0 + \left(\sum_{i=1}^n x_i^3\right)\alpha_1 + \left(\sum_{i=1}^n x_i^4\right)\alpha_2 = \sum_{i=1}^n x_i^2 y_i \end{cases}$$

Stąd $\alpha_0, \alpha_1, \alpha_2$