先端データ解析論 第3回レポート

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大問 1.

$$\mathop{\rm argmax}_z T(z) = \max(0, \theta + u - \lambda) + \min(0, \theta + u + \lambda)$$
を証明する。

証明

$$T(z) = \lambda |z| + u(\theta - z) + \frac{1}{2}$$
 であるので、これを最大化する z は $z > 0$ ならば、

$$\frac{d}{dx}T(z) = \lambda - u - (\theta - z) = \lambda - u - \theta + z = 0$$

$$\therefore z = \theta + u - \lambda$$

z < 0 ならば、

$$\frac{d}{dx}T(z) = -\lambda - u - (\theta - z) = -\lambda - u - \theta + z = 0$$

$$\therefore z = \theta + u + \lambda$$

よって、

$$\underset{z}{\operatorname{argmax}} T(z) = \begin{cases} \theta + u + \lambda & (z < 0) \\ \theta + u - \lambda & (z \ge 0) \end{cases}$$
$$= \max(0, \theta + u - \lambda) + \min(0, \theta + u + \lambda)$$

大問 2.

```
import matplotlib.pyplot as plt
import numpy as np
 \begin{array}{lll} \textbf{def} & \texttt{get\_samples} \, (\, \texttt{x\_samples} \, , & \texttt{f} \, ) \colon \\ & \textbf{return} & \texttt{f} \, (\, \texttt{x\_samples} \, ) \, + \, 0.2 \, * & \texttt{np.random.randn} \, (\, \textbf{len} \, (\, \texttt{x\_samples} \, ) \, ) \end{array} 
\begin{array}{lll} \textbf{def} & \ker \left( \, x \,, \, \, c \,, \, \, h = 0 \,. \, 2 \, \right) \colon \\ & \operatorname{norm} \, = \, x \, - \, \, c \\ & \mathbf{return} & \operatorname{np.exp} \left( - \, \operatorname{norm} \, **2 \, \, / \, \, \left( \, 2 \, \, * \, \, \left( \, h \, **2 \, \right) \, \right) \, \right) \end{array}
def ADM(samples_x , samples_y , lamb=1, h=0.2):
    dim = len(samples_x)
    u, z = np.zeros(dim), np.zeros(dim)
    K = kern_matrix(samples_x , h)
         iteration_cycles = 1500
for i in range(iteration_cycles):
    theta = next_theta(K, samples_y, u, z, lamb, h)
    z = next_z(theta, u, lamb)
    u = next_u(theta, u, z)
return theta
\mathbf{c}
def next_u(theta, u, z):
    return theta + u - z
def kern_model_gen(x_samples, y_samples, lamb=1, h=0.2):
    est_theta = ADM(x_samples, y_samples, lamb, h)
    def _model(x):
        return np.dot(est_theta, kerns(x, x_samples, h))
    v.model = np.vectorize(_model)
    return v_model
np.random.seed()
'- w max = -3, 3
x_min, x_max = n = 50
N = 1000
x = np.linspace(x_min, x_max, n)
X = np.linspace(x_min, x_max, N)
y = org_model(x)
_y = get_samples(x, org_model)
\begin{array}{l} lamb \ = \ 0.1 \\ h \ = \ 0.7 \\ est\_model \ = \ kern\_model\_gen(x, \ \_y \ , \ lamb \ , \ h) \\ Y \ = \ est\_model(X) \end{array}
```

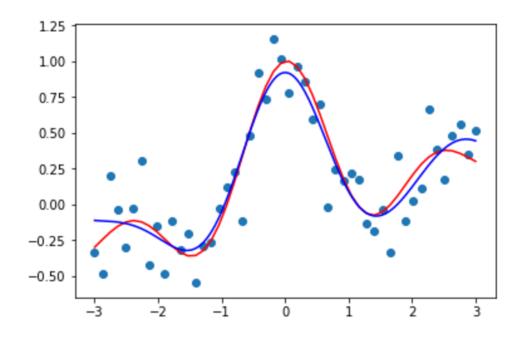


図 1: スパース回帰の結果