

宿題 1

二乗ヒンジ損失に対する適応正則化分類の $\boldsymbol{\mu}$ の解が次式で与えられることを示す。

$$\hat{\boldsymbol{\mu}} = \tilde{\boldsymbol{\mu}} + \frac{y \max(0, 1 - \tilde{\boldsymbol{\mu}}^T \boldsymbol{\phi}(\mathbf{x})y) \tilde{\boldsymbol{\Sigma}} \boldsymbol{\phi}(\mathbf{x})}{\boldsymbol{\phi}(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \boldsymbol{\phi}(\mathbf{x}) + \gamma} \quad (1)$$

損失は,

$$\begin{aligned} J(\boldsymbol{\mu}, \boldsymbol{\Sigma}) = & \left(\max(0, 1 - \boldsymbol{\mu}^T \boldsymbol{\phi}(\mathbf{x})y) \right)^2 + \boldsymbol{\phi}(\mathbf{x})^T \boldsymbol{\Sigma} \boldsymbol{\phi}(\mathbf{x}) \\ & + \gamma \left\{ \log \frac{\det(\tilde{\boldsymbol{\Sigma}})}{\det\{\boldsymbol{\Sigma}\}} + \text{tr}(\tilde{\boldsymbol{\Sigma}}^{-1} \boldsymbol{\Sigma}) + (\boldsymbol{\mu} - \tilde{\boldsymbol{\mu}})^T \tilde{\boldsymbol{\Sigma}}^{-1} (\boldsymbol{\mu} - \tilde{\boldsymbol{\mu}}) - d \right\} \end{aligned} \quad (2)$$

と表される。 f の x による劣微分を $\partial_x f$ などと表すと、この損失の $\boldsymbol{\mu}$ による偏微分は,

$$\frac{\partial J}{\partial \boldsymbol{\mu}} = 2 \cdot \max(0, 1 - \boldsymbol{\mu}^T \boldsymbol{\phi}(\mathbf{x})y) \cdot \partial_{\boldsymbol{\mu}} \max(0, 1 - \boldsymbol{\mu}^T \boldsymbol{\phi}(\mathbf{x})y) + \gamma \left(2 \tilde{\boldsymbol{\Sigma}}^{-1} \boldsymbol{\mu} - 2 \tilde{\boldsymbol{\Sigma}}^{-1} \tilde{\boldsymbol{\mu}} \right) \quad (3)$$

となる。 $z = \boldsymbol{\mu}^T \boldsymbol{\phi}(\mathbf{x})y$ とおくと,

$$\frac{\partial z}{\partial \boldsymbol{\mu}} = y \boldsymbol{\phi}(\mathbf{x}) \quad (4)$$

であり, また,

$$\max(0, 1 - z) = \begin{cases} 1 - z & (z < 1) \\ 0 & (z \geq 1) \end{cases} \quad (5)$$

$$\partial_{\boldsymbol{\mu}} \max(0, 1 - z) = \begin{cases} -y \boldsymbol{\phi}(\mathbf{x}) & (z < 1) \\ \mathbf{0} & (z > 1) \\ [-y \boldsymbol{\phi}(\mathbf{x}), \mathbf{0}] & (z = 1) \end{cases} \quad (6)$$

$$(7)$$

である。

$z > 1$ のとき, 式 (3) - (6) から,

$$\frac{\partial J}{\partial \boldsymbol{\mu}} = 2\gamma \left(\tilde{\boldsymbol{\Sigma}}^{-1} \boldsymbol{\mu} - \tilde{\boldsymbol{\Sigma}}^{-1} \tilde{\boldsymbol{\mu}} \right) = \mathbf{0} \quad (8)$$

よって,

$$\hat{\boldsymbol{\mu}} = \tilde{\boldsymbol{\mu}} \quad (9)$$

$z < 1$ のとき, 式 (3) - (6) から,

$$\frac{\partial J}{\partial \boldsymbol{\mu}} = 2(1 - \boldsymbol{\mu}^T \boldsymbol{\phi}(\mathbf{x})y)(-y \boldsymbol{\phi}(\mathbf{x})) + 2\gamma \left(\tilde{\boldsymbol{\Sigma}}^{-1} \boldsymbol{\mu} - \tilde{\boldsymbol{\Sigma}}^{-1} \tilde{\boldsymbol{\mu}} \right) = \mathbf{0} \quad (10)$$

$\boldsymbol{\mu}^T \phi(\mathbf{x})$ はスカラーなので転置しても値は変わらないことから,

$$(-y\phi(\mathbf{x}))(1-y\phi(\mathbf{x})^T \boldsymbol{\mu}) + \gamma(\tilde{\boldsymbol{\Sigma}}^{-1} \boldsymbol{\mu} - \tilde{\boldsymbol{\Sigma}}^{-1} \tilde{\boldsymbol{\mu}}) = \mathbf{0} \quad (11)$$

$$y^2 \phi(\mathbf{x}) \phi(\mathbf{x})^T \boldsymbol{\mu} - y\phi(\mathbf{x}) + \gamma \tilde{\boldsymbol{\Sigma}}^{-1} \boldsymbol{\mu} - \gamma \tilde{\boldsymbol{\Sigma}}^{-1} \tilde{\boldsymbol{\mu}} = \mathbf{0} \quad (12)$$

$$(\gamma \tilde{\boldsymbol{\Sigma}}^{-1} + y^2 \phi(\mathbf{x}) \phi(\mathbf{x})^T) \boldsymbol{\mu} = \gamma \tilde{\boldsymbol{\Sigma}}^{-1} \tilde{\boldsymbol{\mu}} + y\phi(\mathbf{x}) \quad (13)$$

$$\left(\tilde{\boldsymbol{\Sigma}}^{-1} + \frac{y^2}{\gamma} \phi(\mathbf{x}) \phi(\mathbf{x})^T \right) \boldsymbol{\mu} = \tilde{\boldsymbol{\Sigma}}^{-1} \tilde{\boldsymbol{\mu}} + \frac{y}{\gamma} \phi(\mathbf{x}) \quad (14)$$

いま, $y \in \{-1, 1\}$ を考えるから, $y^2 = 1$ となる。また, ShermanMorrison-Woodbury 公式から,

$$\left(\tilde{\boldsymbol{\Sigma}}^{-1} + \frac{1}{\gamma} \phi(\mathbf{x}) \phi(\mathbf{x})^T \right)^{-1} = \left(\tilde{\boldsymbol{\Sigma}} - \frac{\tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) \phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}}}{\phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) + \gamma} \right) \quad (15)$$

が成立する。よって,

$$\hat{\boldsymbol{\mu}} = \left(\tilde{\boldsymbol{\Sigma}}^{-1} + \frac{1}{\gamma} \phi(\mathbf{x}) \phi(\mathbf{x})^T \right)^{-1} \left(\tilde{\boldsymbol{\Sigma}}^{-1} \tilde{\boldsymbol{\mu}} + \frac{y}{\gamma} \phi(\mathbf{x}) \right) \quad (16)$$

$$= \left(\tilde{\boldsymbol{\Sigma}} - \frac{\tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) \phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}}}{\phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) + \gamma} \right) \left(\tilde{\boldsymbol{\Sigma}}^{-1} \tilde{\boldsymbol{\mu}} + \frac{y}{\gamma} \phi(\mathbf{x}) \right) \quad (17)$$

$$= \tilde{\boldsymbol{\mu}} - \frac{\tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) \phi(\mathbf{x})^T}{\phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) + \gamma} \tilde{\boldsymbol{\mu}} + \frac{y}{\gamma} \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) \left(1 - \frac{\phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x})}{\phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) + \gamma} \right) \quad (18)$$

$$= \tilde{\boldsymbol{\mu}} - \frac{\tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) \phi(\mathbf{x})^T}{\phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) + \gamma} \tilde{\boldsymbol{\mu}} + \frac{y \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x})}{\phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) + \gamma} \quad (19)$$

$$= \tilde{\boldsymbol{\mu}} + \frac{\tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x})}{\phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) + \gamma} (y - \phi(\mathbf{x})^T \tilde{\boldsymbol{\mu}}) \quad (20)$$

$y \in \{-1, 1\}$ から,

$$\hat{\boldsymbol{\mu}} = \tilde{\boldsymbol{\mu}} + \frac{y \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x})}{\phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) + \gamma} (1 - \phi(\mathbf{x})^T \tilde{\boldsymbol{\mu}} y) \quad (21)$$

$$= \tilde{\boldsymbol{\mu}} + \frac{y(1 - \tilde{\boldsymbol{\mu}}^T \phi(\mathbf{x}) y)}{\phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) + \gamma} \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) \quad (22)$$

$z = 1$ のときの劣勾配を $\mathbf{0}$ としてしまえば, 式 (9), (22) をまとめて,

$$\hat{\boldsymbol{\mu}} = \tilde{\boldsymbol{\mu}} + \frac{y \max(0, 1 - \tilde{\boldsymbol{\mu}}^T \phi(\mathbf{x}) y)}{\phi(\mathbf{x})^T \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) + \gamma} \tilde{\boldsymbol{\Sigma}} \phi(\mathbf{x}) \quad (23)$$