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A Since in this case we are dealing with gaussian distributions, for mean  $\bar{\phi}$  & variance  $\sigma^2$

$$p.d.f = f(y|\bar{\phi}, \sigma^2) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(y-\bar{\phi})^2}{2\sigma^2}\right)$$

For given  $y$ ,

$$\begin{aligned} \text{Likelihood of } \phi \text{ given } y & \text{ is } L(\phi|y, \sigma^2) \\ &= \frac{1}{\sqrt{2\pi}\sigma} \exp\left[-\frac{(y-\phi)^2}{2\sigma^2}\right] \end{aligned}$$

We only consider one value of  $y$  & thus the variance is also constant.

$$\begin{aligned} \Rightarrow \text{negative log likelihood} &= -\log(L(\phi|y, \sigma^2)) \\ &= -\log\left(\frac{1}{\sqrt{2\pi}\sigma} \exp\left[-\frac{(y-\phi)^2}{2\sigma^2}\right]\right) \\ &= -\left(-\frac{1}{2}\log(2\pi\sigma^2) - \frac{(y-\phi)^2}{2\sigma^2}\right) \\ &= C + \frac{(y-\phi)^2}{2\sigma^2} \end{aligned}$$

$\therefore$   $C$  &  $\sigma$  are constants, so to minimize negative of the loglikelihood, we need to minimize  $(y-\phi)^2$

For the convenience of it, we will keep  
the  $\frac{1}{2}$

$$\Rightarrow \boxed{\min : \frac{1}{2} (y - \phi)^2 \quad \forall y \in Y.}$$

To introduce a penalty function, we add  
the absolute value of  $\phi$  multiplied by  
a multiplier,

$$\forall \quad \boxed{\min : \frac{1}{2} (y - \phi)^2 + \lambda |\phi|}$$

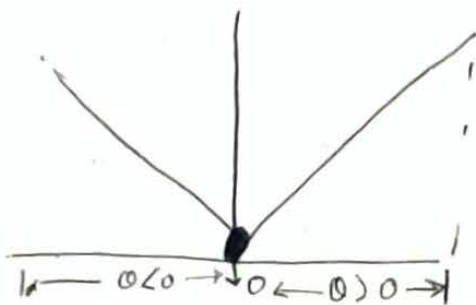
$$\Rightarrow \boxed{\delta_{\lambda}(y) = \arg \min_{\phi} \frac{1}{2} (y - \phi)^2 + \lambda |\phi|} \quad \text{--- (1)}$$

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### MINIMIZING

To minimize, we differentiate equation (1) with  
respect to  $\phi$  & set it to zero,

checking  $|\phi|$  is not differentiable, so  
we divide it in differentiable zones.



$$\frac{d S_p(y)}{d \phi} = \frac{d}{d \phi} \left( \frac{1}{2} (y - \phi)^2 + \lambda |\phi| \right)$$

$$= -(y - \phi) + \lambda \cdot \text{sign}(\phi)$$

①

$$\boxed{\phi > 0}$$

$$\Rightarrow \frac{d S_p}{d \phi} = -(y - \phi) + \lambda = 0 \Rightarrow \phi = y - \lambda$$

$$\text{and } \phi > 0$$

$$\Rightarrow y - \lambda > 0 \Rightarrow y > \lambda$$

②

$$\boxed{\phi < 0}$$

$$\Rightarrow \frac{d S_p}{d \phi} = -y + \phi - \lambda = 0 \Rightarrow \phi = y + \lambda$$

$$\text{and } \phi < 0$$

$$\Rightarrow y < -\lambda$$

$$\textcircled{3} \cdot \phi = 0$$

$$\Rightarrow -\lambda < y < \lambda$$

$J_{\text{norm}}$

① and ②

$$y > \lambda$$

$$\text{and } y < -\lambda$$

$$\Rightarrow |y| > \lambda$$

$$\text{and } J_{\text{norm}} |y| < \lambda \text{ then } \phi = 0.$$

$\Rightarrow$

$$S_p(y) = \text{sign}(y) \cdot (|y| - \lambda) +$$