

## WEEKLY PARTICIPATION 1

Recall that we setup the problem of learning a linear separator  $y = \text{sign}(\langle \mathbf{w}, \mathbf{x} \rangle)$  for binary classification as follows (SVM):

$$\mathbf{w}_* = \arg \min_{\mathbf{w}} \frac{1}{n} \sum_{i=1}^n \phi(y^{(i)} \langle \mathbf{w}, \mathbf{x}^{(i)} \rangle) + \frac{\lambda}{2} \|\mathbf{w}\|_2^2,$$

to achieve the following goals:

- Encourage the prediction to have the correct sign on all the training data, and be confident,  $y^{(i)} \langle \mathbf{w}, \mathbf{x}^{(i)} \rangle \geq 1$ . Recall that the specific value of 1 is not important, just that it is encouraged that  $y^{(i)} \langle \mathbf{w}, \mathbf{x}^{(i)} \rangle$  be bounded away from zero for all the training data.
- Encourage the model to be stable, i.e.  $\|\mathbf{w}\|_2$  to be small.

We chose  $\phi$  to be the hinge loss  $\phi(t) = \max\{1 - t, 0\}$  because it encourages the first point to be true. Which of the following choices similarly encourage the first property?

- (A)  $\phi(t) = \log(1 + \exp(-t))$
- (B)  $\phi(t) = \exp(-t)$
- (C)  $\phi(t) = 1 - \tanh(t)$

Answer by listing none, one, two, or all of these options.