## CMPUT 466 W6

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### Problem 1

say there are 90 positive samples out of 100. Then if we always predict t=1,

- 90 of the samples are true positive (TP)
- 10 of the samples are false positive (FP)
- 0 of the samples are false negative (FN)

$$P = \frac{TP}{TP + FP} = \frac{90}{90 + 10} = 0.9$$

$$R = \frac{TP}{TP + FN} = \frac{90}{90 + 0} = 1$$

$$F_1 = \frac{2PR}{P + R} = \frac{2 \times 0.9 \times 1}{0.9 + 1} = \frac{18}{19} \approx 0.9474$$

The issue here is that the  $F_1$  score is insensitive. With a trivial classifier, we achieved an  $F_1$  score of 94.7%.

The positive category should be the minority class. This will have a more meaningful result. i.e., we change the meaning of positive samples to be t=0 and negative to be t=1

### Problem 2

Prove 
$$\sigma(-z) = 1 - \sigma(z)$$

Proof.

$$\sigma(z) = \frac{1}{1 + e^{-z}}$$
$$1 - \sigma(z) = 1 - \frac{1}{1 + e^{-z}}$$
$$= \frac{1 + e^{-z}}{1 + e^{-z}} - \frac{1}{1 + e^{-z}}$$

$$= \frac{1}{e^{z}(1 + e^{-z})}$$
$$= \frac{1}{1 + e^{-(-z)}}$$
$$= \sigma(-z)$$

Therefore, we have proven that  $\sigma(z)$  is symmetric about the point (0,0.5)

# Problem 3

Prove minimizing  $KL(\mathbf{t}||\mathbf{y}) \Leftrightarrow \text{minimizing } J = -t\log y - (1-t)\log(1-y)$ 

$$\begin{aligned} \textit{Proof.} & \text{ minimize } \sum_{k=1}^K t_k \log \frac{t_k}{y_k} \\ &\Leftrightarrow \text{ minimize } \sum_{k=1}^K t_k \log t_k - \sum_{k=1}^K t_k \log y_k \\ &\Leftrightarrow \text{ minimize } - \sum_{k=1}^K t_k \log y_k \\ &\Leftrightarrow \text{ minimize } - (t \log y + (1-t) \log (1-y)) \\ &\Leftrightarrow \text{ minimize } J = -t \log y - (1-t) \log (1-y) \end{aligned}$$