

## Semi-supervised learning: loss fn

Salimans et al 2016 describe approach to loss fns.

If you have a  $N$  way softmax classifier, then use a

$N+1$  classifier to allow for a fake image. Then

note only need a  $N$  way classifier - subtracting a constant does not change softmax probabilities

Output neurons are  $x_1, x_2, x_3$ .

$$\text{Probability} = \frac{e^{x_i}}{\sum_j e^{x_j}}$$

subtract  $a$ , so have  $x_1 - a, x_2 - a, x_3 - a$

$$\text{gives } \frac{e^{x_i - a}}{\sum_j e^{x_j - a}} = \frac{\cancel{e^{x_i}} \cancel{e^{-a}}}{\cancel{e^{-a}} \sum_j e^{x_j}} = \frac{e^{x_i}}{\sum_j e^{x_j}}$$

So set  $y_1 = x_1 - x_3, y_2 = x_2 - x_3, y_3 = 0$ . Therefore  $N+1$  neurons hardwired to 0.

$$P(\text{image fake}) = \frac{e^0}{\sum_{j=1}^N e^{y_j} + e^0} = \frac{1}{\sum_{j=1}^N e^{y_j} + 1}$$

$$P(\text{image real}) = \frac{\sum e^{y_j}}{\sum e^{y_j} + 1}$$

$$\begin{aligned} \text{Loss for a real image} &= -\log(\text{prob real}) \\ &= -[\sum e^{x_i} - \log(\sum e^{x_i} + 1)] \\ &= -\log \sum e^{x_i} + \log(\sum e^{x_i} + 1) \end{aligned}$$

$$\begin{aligned} \text{Loss for a fake image} &= -\log(\text{prob fake}) \\ &= -[\log 1 - \log(\sum e^{x_i} + 1)] \\ &= \log(\sum e^{x_i} + 1) \end{aligned}$$

1) Loss is the cross entropy loss.

2) For a mini-batch, sum the loss.