

$$\text{Sigmoid} = \sigma(x) = \frac{1}{1+e^{-x}}$$

$$\sigma(0) = \frac{1}{2}$$

$$\frac{d}{dx} \text{ of Sigmoid} = \sigma'(x) = \sigma(x)(1-\sigma(x))$$

$$\sigma'(0) = \frac{1}{2} \times \frac{1}{2} = \frac{1}{4}$$

$$\text{Maclaurin} = f(\epsilon) = f(0) + f'(0)\epsilon$$

$$\sigma'(\epsilon) = \frac{1}{4}$$

$$\text{Tanh} = \frac{d}{dx} \tanh x = 1 - \tanh^2 x$$

$$\tanh(0) = 0$$

$$\frac{d}{dx} - \tanh'(0) = 1 - \tanh^2(0) = 1$$

$$\text{Maclaurin} - f(\epsilon) = f(0) + f'(0)\epsilon$$

$$\tanh(\epsilon) \approx \epsilon$$

$$\frac{d}{dx} \text{ Mac} - \tanh'(\epsilon) \approx 1$$

$$2. \quad \sigma_{\text{approx}}(\epsilon) \leq \tanh_{\text{approx}}(\epsilon)$$

$$\frac{1}{2} + \frac{1}{4}\epsilon \leq \epsilon$$

$$\frac{1}{2} \leq \epsilon - \frac{1}{4}\epsilon$$

$$\frac{1}{2} \leq \frac{3}{4}\epsilon$$

$$\epsilon \geq \frac{2}{3}$$

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$$\frac{1}{2} + \frac{1}{4}\epsilon \leq \epsilon \Rightarrow \epsilon \geq \frac{2}{3}$$

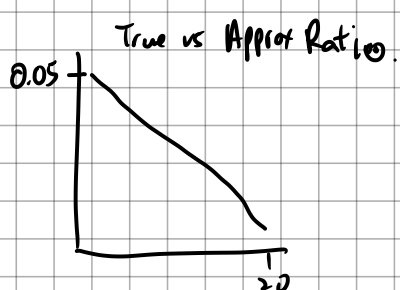
$$\epsilon \in \left[\frac{2}{3}, \infty\right)$$

3. Both approx ratio  $p(\epsilon)^T$  & true ratio  $p(\epsilon)^T$  exhibit exponential decay as  $T$  increase.

$$\text{True ratio } p(\epsilon)^T = \left( \frac{\sigma(\epsilon)}{\tanh(\epsilon)} \right)^T$$

$$\text{Approximate ratio: } p_{\text{approx}}(\epsilon)^T = (0.25)^T$$

$$\epsilon = 0.05, T = 1 \text{ to } 20$$



4.

Because backprop through time multiplies the derivatives across many steps, sigmoid causes gradient to vanish rapidly.

But tanh with derivative close to 1 will make gradients much longer.

$\therefore$  Tanh slows down vanish-gradient problem & allows RNNs to learn dependencies over more time steps