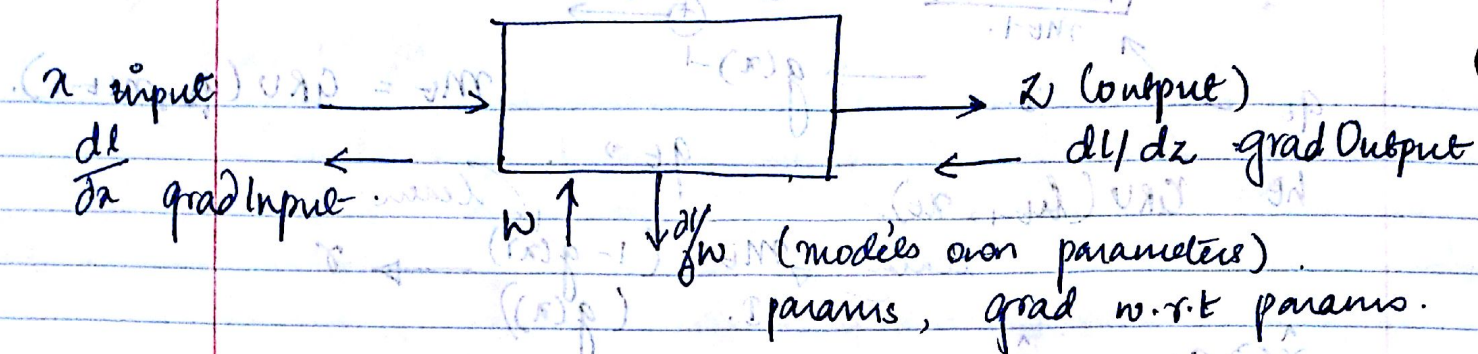


TORCH: understanding backward & forward pass.



$z = \text{model} : \text{forward}(x)$. \rightarrow calls $\text{updateOutput}(x)$ internally.

$\frac{dl}{dz} = \text{model} : \text{backward}(x, \frac{\partial l}{\partial z})$

\rightarrow update Grad Input ($x, \frac{\partial l}{\partial z}$) \rightarrow the $\frac{\partial l}{\partial x}$.
 \rightarrow acc Grad Param = computes gradient w.r.t model's own params (w) "trainable weights"

you can zero this acc $\frac{dl}{dw}$:
 $\text{zeroGradParams}()$

you can update $w^* = w - lr * \frac{\partial l}{\partial w}$ params

$$w^* = w - lr \times \frac{\partial l}{\partial w}$$

$w, dl-dw = \text{model} : \text{getParams}()$

$\text{feval}(w_{\text{new}})$

\rightarrow learnable params.

if $w \sim w_{\text{new}}$

// copy weights new

$w : \text{copy}(w_{\text{new}})$

end

$dl-dw : \text{zero}$ or

$\text{zeroGradParams}()$

0 the $\frac{dl}{dw}$.

$x \leftarrow \frac{\partial p}{\partial p}$ new batch.

$(x) \text{ prediction} = \text{model} : \text{forward}(x)$

$\text{loss} = \text{criterion} : \text{forward}(\text{pred}, \text{target})$

$\frac{dl}{dz} = \text{model} : \text{backward}(\text{criterion})$

$\text{model} : \text{backward} \rightarrow$ gives $\frac{dl}{dz}, \frac{dl}{dw}$

Return $\text{loss}, \frac{dl}{dw}$