# TTIC 31230, Fundamentals of Deep Learning 

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Backpropagation with Arrays and Tensors

## Program Values as Objects

In a framework the program (or deep model) variables are objects in the sense of object oriented programming or Python.

Each object $x$ stores its input objects in its instance variables and has an instance variable $x$.value storing its value.

The instance variable $x$.value is filled by sending $x$ a forward message after its inputs have computed their values.
Each object $x$ has an instance variable $x$.grad storing $\partial \mathcal{L} / \partial x$.
$x$.grad is filled by the backward methods of objects $y$ that use $x$ as an input. The backward method for $y$ is called after $y$.grad has been filled and adds into $x$.grad for each input $x$.

## Scalar Products

Consider a scalar product $z=x y$.
The forward method for $z$ computes.

$$
z . \text { value }=x . \text { value } * y . \text { value }
$$

The backward method for $z$ computes

$$
\begin{aligned}
& x \cdot \operatorname{grad}+=z \cdot \operatorname{grad} * y \cdot \text { value } \\
& y \cdot \operatorname{grad}+=z \cdot \operatorname{grad} * x \cdot \text { value }
\end{aligned}
$$

## Handling Arrays

Consider an inner product between vectors

$$
z=x^{\top} y
$$

In this case $z$.forward does

$$
\begin{gathered}
z \cdot \text { value }=0 \\
\text { for } i \quad z \text {.value }+=x \cdot \text { value }[i] * y \cdot \text { value }[i]
\end{gathered}
$$

The backward method for $z$ treats each $+=$ instruction separately and does.

$$
\begin{array}{ll}
\text { for } i & x \cdot \operatorname{grad}[i]+=y \cdot \text { value }[i] * z \cdot \operatorname{grad} \\
\text { for } i & y \cdot \operatorname{grad}[i]+=x \cdot \operatorname{value}[i] * z \cdot \operatorname{grad}
\end{array}
$$

## Handling Arrays

Now consider multiplying a vector $x$ by a matrix $W$.

$$
y=W x
$$

In this case case $y$.forward does

$$
\begin{gathered}
\text { for } j \quad y \text {.value }[j]=0 \\
\text { for } i, j \quad y \text {.value }[j]+=W \cdot \operatorname{value}[j, i] * x \text {.value }[i]
\end{gathered}
$$

The backward procedure $y$.backward treats each individual $+=$ as a scalar product and does

$$
\begin{array}{ll}
\text { for } i, j & x \cdot \operatorname{grad}[i]+=W \cdot \text { value }[j, i] * y \cdot \operatorname{grad}[j] \\
\text { for } i, j & W \cdot \operatorname{grad}[j, i]+=x \cdot \operatorname{value}[i] * y \cdot \operatorname{grad}[j]
\end{array}
$$

## A Linear Threshold Layer

$$
\begin{aligned}
& \quad s=\sigma(W h-B) \\
& \text { for } j \quad \tilde{s}[j]=0 \\
& \text { for } j, i \tilde{s}[j]+=W[j, i] h[i] \\
& \text { for } j \quad s[j]=\sigma(\tilde{s}[j]-B[j])
\end{aligned}
$$

Backpropagation is also done with loops treating each individual assignment and += instruction.

## General Tensor Operations

In practice all deep learning source code can be written using scalar assignments and loops over scalar assignments. For example:

$$
\text { for } h, i, j, k Y[h, i, j]+=A[h, i, k] B[h, j, k]
$$

has backpropagation loops
for $h, i, j, k \quad A \cdot \operatorname{grad}[h, i, k]+=Y \cdot \operatorname{grad}[h, i, j] B \cdot \operatorname{value}[h, j, k]$ for $h, i, j, k B \cdot \operatorname{grad}[h, j, k]+=Y \cdot \operatorname{grad}[h, i, j] A$.value $[h, i, k]$

END

