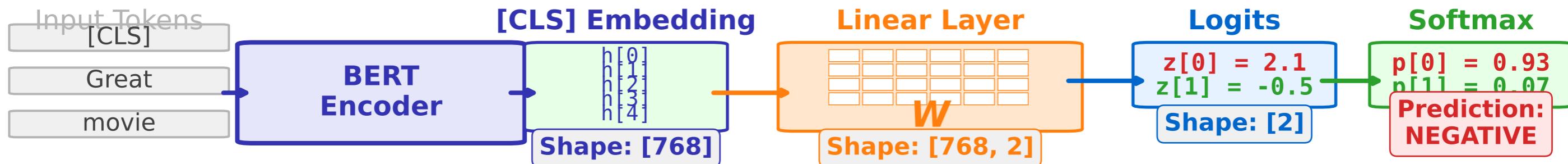


Stage 2: Adding Classifier Head to Pre-trained BERT



Matrix Multiplication Details

Linear Layer Computation:

$$z = W^T \cdot h + b$$

Where:

h = [CLS] embedding (768 dimensions)
 W = Weight matrix (768×2)
 b = Bias vector (2 dimensions)
 z = Output logits (2 dimensions)

Example (simplified to 3D):

$$\begin{aligned} h &= [0.5, -0.2, 0.8]^T \\ W &= [[0.3, -0.1], \\ &\quad [0.2, 0.4], \\ &\quad [-0.1, 0.5]] \\ b &= [0.1, -0.05] \\ z &= [0.3*0.5 + 0.2*(-0.2) + (-0.1)*0.8, \\ &\quad -0.1*0.5 + 0.4*(-0.2) + 0.5*0.8] + b \\ &= [0.03, 0.27] + [0.1, -0.05] \\ &= [0.13, 0.22] \end{aligned}$$

Initialization Strategy

Classifier Head Initialization:

BERT Layers (Stage 1):

- Load pre-trained weights
- Already optimized on Wikipedia
- Frozen or fine-tuned slowly

Linear Layer (Stage 2):

- Random initialization
- Xavier/Glorot uniform:
 $W \sim U(-\sqrt{6/(768+2)}, \sqrt{6/(768+2)})$
- Bias initialized to zeros: $b = [0, 0]$

Why Random Init?:

- No prior knowledge of task
- Fine-tuning will adapt to sentiment
- Fast convergence (3-5 epochs)