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# Karpathy Inference for Llama-2

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## Abstract

Step-by-step re-implementation of Karpathy code for llama2.

## 1 XXX

### 1.1 Conventions

If  $T$  is a table of dimensions  $n_1 \times n_2$ , then  $T_i$  is a vector of length  $n_2$

### 1.2 Glossary

Abbreviation	C code	Comments
$n_D$	dim	
$n_{HD}$	hidden_dim	
$n_{KVD}$	n_kv_dim	$n_D \times n_{KVH} / n_H$
$n_L$	n_layers	
$n_H$	n_heads	
$n_{KVH}$	n_kv_heads	
$n_V$	vocab_size	
$n_S$	seq_len	
$s_H$	head_size	
$n'_D$	ispc_dim	

Table 1: Scalars: Mapping math notation to C code

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Purpose	Abbreviation	C code	Dim1	Dim2	Dim3
	$x$	x	$n_D$		
	$xb$	xb	$n_D$		
	$xb2$	xb2	$n_D$		
	$xhb$	xhb	$n_{HD}$		
	$xhb2$	xhb2	$n_{HD}$		
	$q$	q	$n_H$	$s_H$	
	$kc$	key_cache	$n_L$	$n_S$	$n_{KVD}$
	$vc$	val_cache	$n_L$	$n_S$	$n_{KVD}$
	$att$	att	$n_H$	$n_S$	
	$logits$	logits	$n_V$		

Table 2: State: Mapping math notation to C code

Purpose	Abbreviation	C code	Dim1	Dim2	Dim3
Token Embedding	$W_t$	token embedding table	$n_V$	$n_D$	
Normalization	$W_{att}$	rms_att_weight	$n_L$	$n_D$	
Normalization	$W_{ffn}$	rms_ffn_weight	$n_L$	$n_D$	
XXX	$W_{fin}$	rms_final_weight	$n_D$		
Attention	$W_q$	w_q	$n_L$	$n_D$	$n_H \times s_H$
Attention	$W_k$	w_k	$n_L$	$n_D$	$n_{KVH} \times s_H$
Attention	$W_v$	w_v	$n_L$	$n_D$	$n_{KVH} \times s_H$
Attention	$W_o$	w_o	$n_L$	$n_H \times s_H$	$n_D$

Table 3: Weights: Mapping math notation to C code

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## **2 Utilities**

### 3 Forward

Arguments in Table 4

Argument	Type	Comments
$T$	Transformer	pointer
$t$	integer	token $0 \leq t < n_V$
$p$	integer	pos
Argument	Type	Comments
$d$	integer	dim

Table 4: Arguments/Convenience Variables for forward

Copy token into  $x$ , Equation 1

$$x \leftarrow W_t[t] \quad (1)$$

**for** each layer  $l$  **do** the following

1. attention rmsnorm, Equation 2

$$xb \leftarrow \text{rmsnorm}(x, W_{ra}[l]) \quad (2)$$

2. qkv matmuls for this position, Equations 3, 4, 5.

$$S_{wq}[l][d] \leftarrow \text{matmul}(xb, W_{wq}[l], n_D, n_D) \quad (3)$$

$$S_{kc}[l][p] \leftarrow \text{matmul}(xb, W_{wk}[l], n_D, n_{KVD}) \quad (4)$$

$$S_{vc}[l][p] \leftarrow \text{matmul}(xb, W_{wv}[l], n_D, n_{KVD}) \quad (5)$$

3. residual connection back into  $x$ , Equation 6

$$x \leftarrow x + xb2 \quad (6)$$

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### 3.1 rmsnorm

Input is in Table 5. Output is  $o$ , float vector of length  $n$

Argument	Type
$x$	float vector of length $n$
$w$	float vector of length $n$

Table 5: Arguments for rmsnorm

1.  $\alpha = ((\sum_i x_i^2)/n) + \epsilon$
2.  $o_i \leftarrow \frac{w_i \times x_i}{\alpha}$

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## 3.2 softmax

Arguments in Table 6

Argument	Type
$x$	float vector of length $n$
$n$	integer

Table 6: Arguments for softmax

1.  $m = \max_{i=0}^{i=n-1} x_i$
2.  $\forall_{i=0}^{i=n-1} x_i \leftarrow e^{x_i - m}$
3.  $s = \sum_{i=0}^{i=n-1} x_i$
4.  $\forall_{i=0}^{i=n-1} x_i \leftarrow \frac{x_i}{s}$