

Architecture neuronale pour données textuelles

Janvier 22, 2025

Familles de modèles classiques en apprentissage supervisé

- Modèles de régression pénalisée
- Arbres de décision
- Forêts aléatoires et gradient boosting
- Réseaux de neurones profonds (ou pas)
- Support vector machines (modèles à marge de séparation maximale)

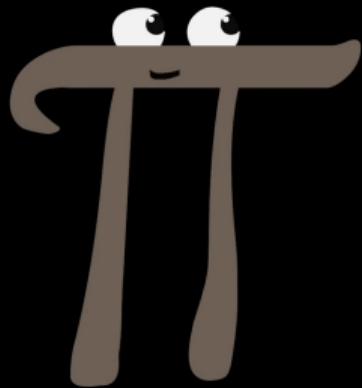
Nous avons besoin d'un vecteur de variables explicatives $X \in \bigtimes_{j=1}^d \mathcal{X}_j$.

Révolution du mécanisme d'attention en 2018

Équivalente à la révolution des couches de convolution en classification d'images
(AlexNet , concours du 30 septembre 2012 au Canada sur la base de données
ImageNet).

- Tokenizer : l'idée est de garder entier les mots très utilisés, mais de découper les mots transformés par la grammaire
- Embedding ou plongement : transformer nos jetons (numérotation) en vecteur dans \mathbb{R}^d
- Architecture neuronale dite **transformer**

Generative Pre-trained Transformer



Generative

Pre-trained

Transformer

The most effective way to learn computer science is to actively engage with the material, practice regularly, and seek help when needed. Here are some specific steps you

Generative

Pre-trained

Transformer

it for a fact, years ago, in the quiet corner in Soho, that this precious brother had spent her money and left her! He was saying the affectionate word, however, with a far more grudging condescension and patronage than he could have shown if their relative merits and positions had

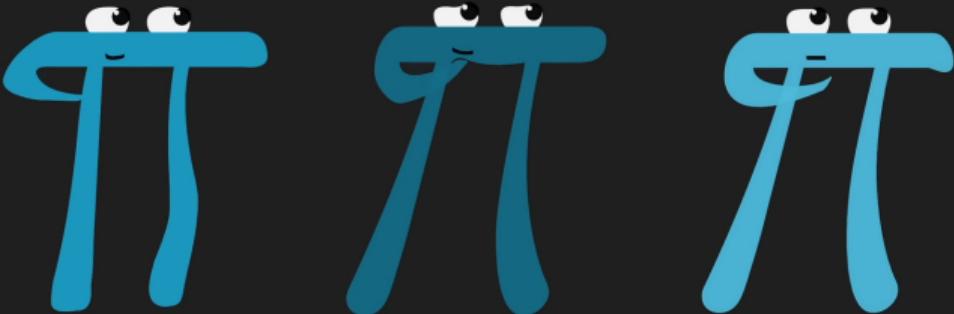


Weights

What defines the model

+1.0	+4.3	+2.0	+0.9	-1.5	+2.9	-1.2	+7.8
+9.2	-2.3	+5.8	+0.6	+1.3	+8.4	-8.5	-8.2
-9.5	+6.6	+5.5	+7.3	+9.5	+5.9	-0.8	+5.6
-7.6	+2.8	-7.1	+8.8	+0.4	-1.7	-4.7	+5.4
-0.9	+1.4	-9.5	+2.3	+2.2	+2.3	+8.8	+3.6
-2.8	-1.2	+3.9	-8.7	+3.3	+3.4	-5.7	-7.3

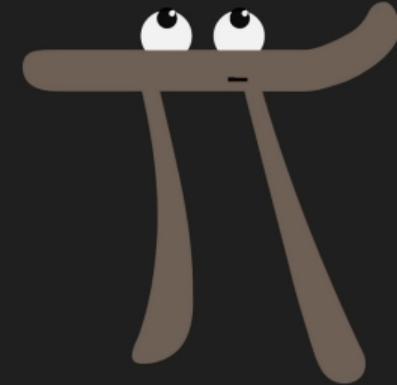
+1.7	-9.5	+6.5	-9.8	+3.5	-4.6	+4.7	+9.2
-5.0	+1.5	+1.8	+1.4	-5.5	+9.0	-1.0	+6.9
+3.9	-4.0	+6.2	-2.0	+7.5	+1.6	+7.6	+3.8
+4.5	+0.0	+9.0	+2.9	-1.5	+2.1	-9.5	-3.9
+3.2	-4.2	+2.3	-1.4	-7.2	-4.0	+1.4	+1.8
+1.5	+3.0	+3.0	-1.4	+7.9	-2.6	-1.3	+7.8



Data

What the model processes

-6.2	-3.7	+0.7	-3.6	-6.2	+0.4	+7.2	-7.0
+8.8	+3.9	+7.9	-2.3	+9.0	-9.4	-7.6	-0.2
+4.7	-2.4	+9.7	+1.7	+3.7	-5.8	+0.3	-2.9
-0.2	-6.3	-5.6	+6.6	-5.6	-1.5	-7.3	+8.7
-5.4	-9.4	+3.2	+2.6	+8.9	-2.5	+4.3	+5.3
-4.9	-8.6	-4.7	+7.4	+4.6	-0.7	-2.1	+4.9
-8.8	+3.6	-9.5	-4.5	-4.9	-4.4	+1.3	+8.0
-1.3	-0.9	+5.1	+5.9	-5.7	+1.7	-6.3	-8.2



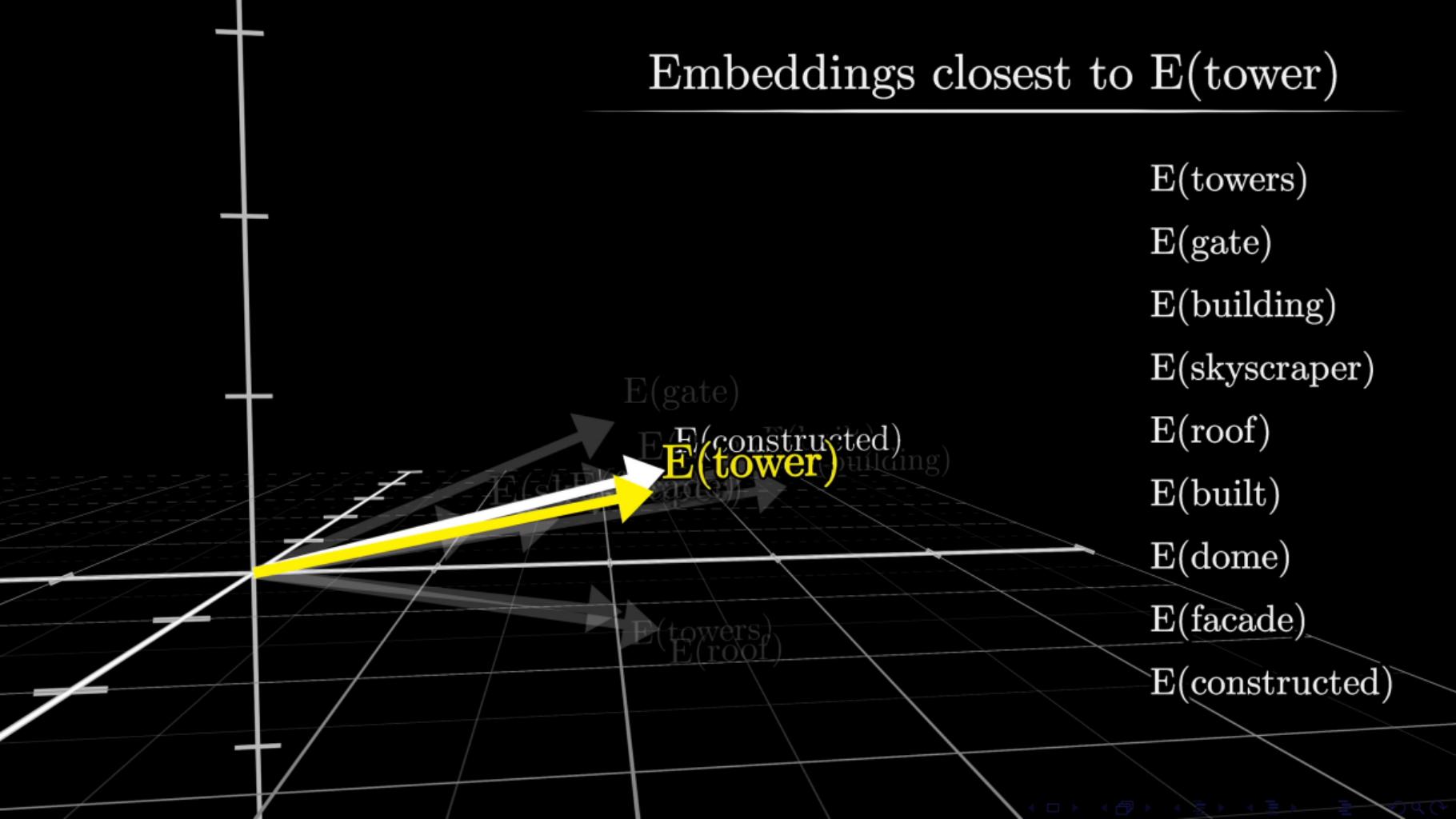
Total parameters = $12,288 \times 50,257 = 617,558,016$

50,257 tokens

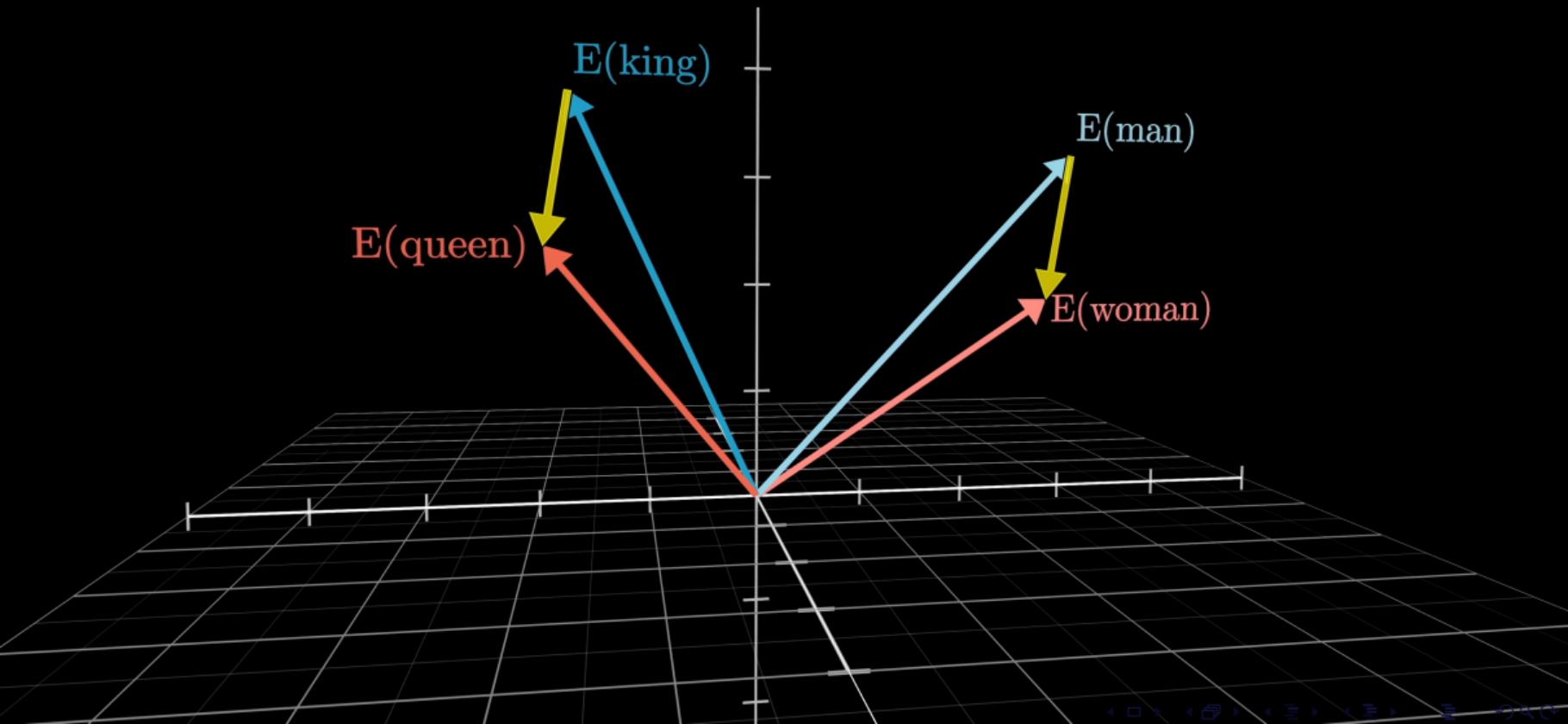
	~	*	*	*	*	*	*	*	*	*	*	*	*	*	*	*	Compar	amplification	ominated	regress	Collider	informants	gazed	<endoftext>
12,288	-4.0	+1.0	+8.5	+0.4	-4.6	+7.5	-2.5	-9.9	-5.0	-3.6	...	+7.1	-0.8	-1.1	-3.2	+7.5	+8.8	+9.7	-2.4	+9.2	+5.8			
	+3.5	-5.1	-5.6	-6.6	+8.4	-4.1	-0.9	-0.1	+5.5	+6.8	...	-7.1	-1.4	+6.8	+6.3	-7.9	-6.8	-3.9	-8.4	-1.5	-7.8			
	+1.4	-5.0	+1.9	-7.6	+9.4	+8.6	-2.1	-5.1	-4.9	-0.3	...	-9.1	+2.8	-1.8	-2.4	+6.1	+4.1	+9.0	-2.9	+7.9	+5.3			
	-2.8	+2.4	-4.2	+7.4	-7.7	-5.7	-6.3	-1.9	+4.9	+0.5	...	-0.2	-9.9	-1.5	-8.6	-5.8	+8.6	-5.6	+7.1	+6.0	-6.7			
	+2.1	-7.6	+4.5	+2.7	+6.2	-0.4	+8.2	-8.9	-4.1	+4.3	...	-1.6	-6.5	-7.8	+6.3	-0.5	+7.6	+4.6	-1.8	-2.5	+0.3			
	+7.7	+4.7	-9.8	+3.8	+8.3	+4.2	-6.4	-0.3	-7.1	-2.8	...	+8.7	+8.4	-4.3	-3.2	+2.0	+9.2	-7.0	-4.8	+7.4	-0.2			
	+7.9	-6.2	+0.6	-3.4	-3.6	-1.1	-1.3	-2.8	+8.2	+4.6	...	+4.5	-4.2	+1.5	+5.5	+5.9	-3.1	+5.4	+4.7	-7.1	+7.2			
	-1.2	-0.3	-1.0	+1.3	+2.4	+0.0	+7.3	+2.5	-2.0	-1.6	...	+6.2	-3.0	-5.7	-8.7	+7.4	+8.3	-7.5	-3.3	-6.4	-7.6			
			
	+7.9	-8.8	+9.5	-8.0	+7.2	+1.3	-2.6	-3.1	+5.1	-3.7	...	+3.1	+0.3	-0.3	+7.9	+1.1	+6.5	+4.5	-9.1	+5.4	-5.6			

W_E = Embedding matrix

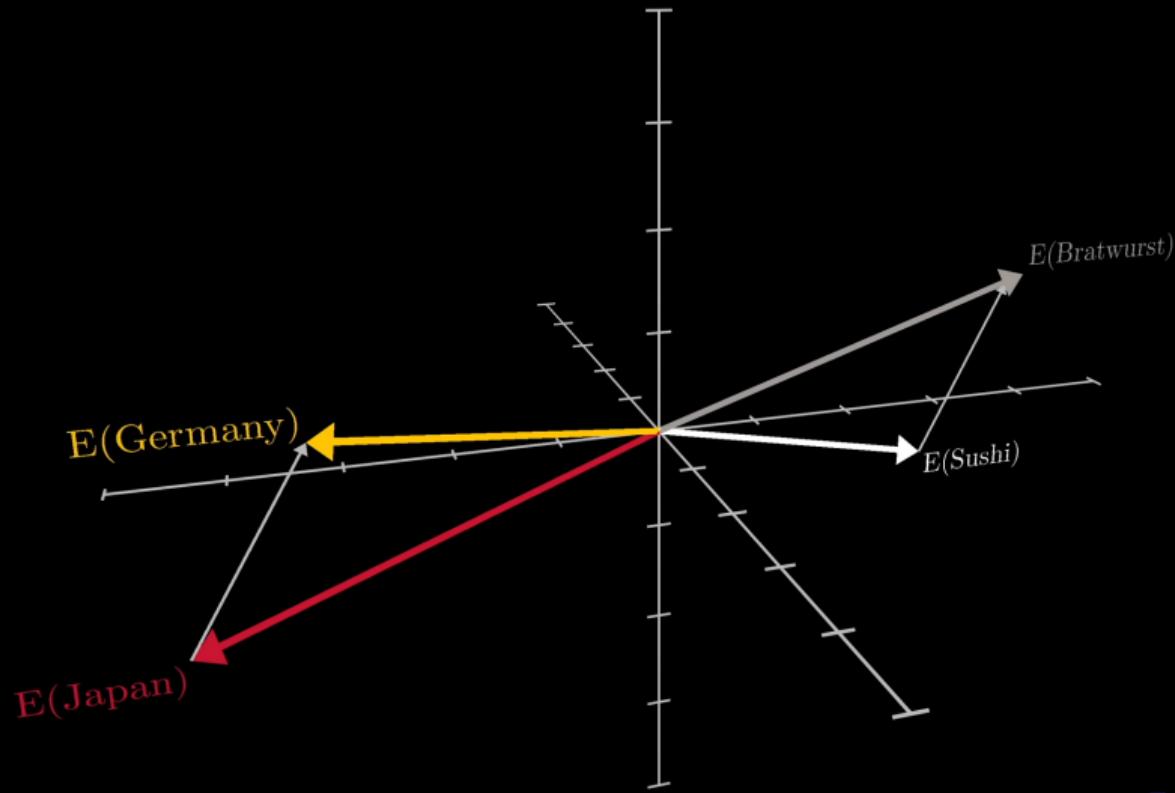
Embeddings closest to $E(\text{tower})$



$$E(\text{queen}) - E(\text{king}) \approx E(\text{woman}) - E(\text{man})$$



$$E(\text{Sushi}) + E(\text{Germany}) - E(\text{Japan}) \approx \boxed{E(\text{Bratwurst})}$$

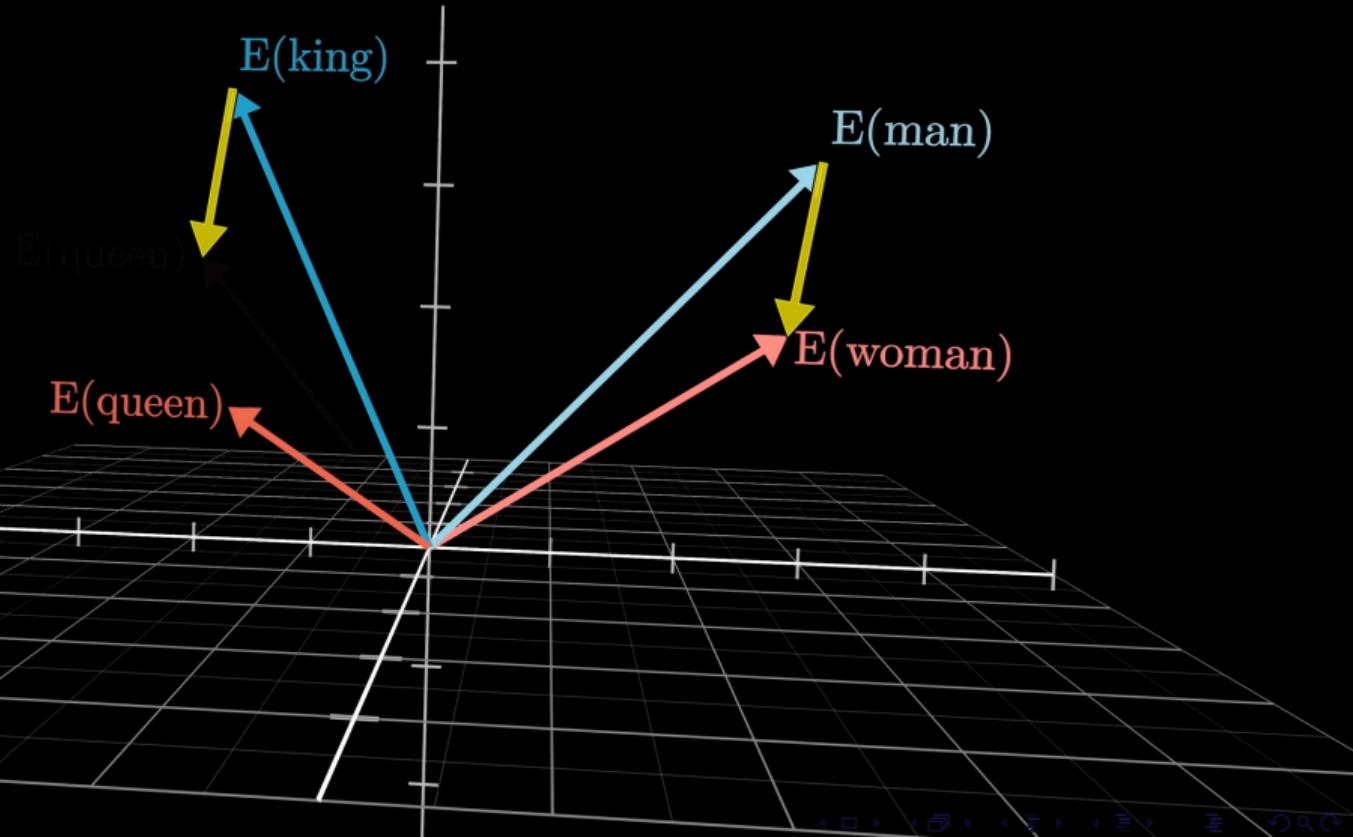
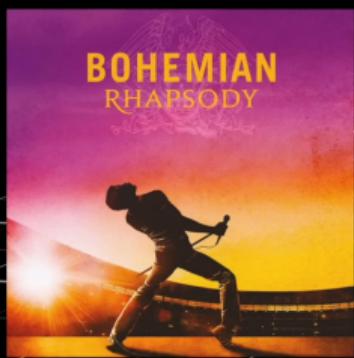
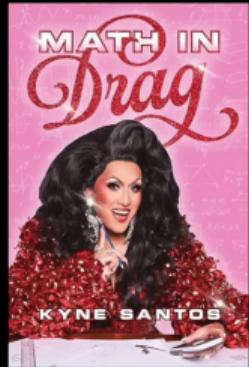


$$\overrightarrow{\text{plur}} := E(\text{cats}) - E(\text{cat})$$

$$\{ \overrightarrow{\text{plur}} \cdot E(\text{puppies}) = 1.85$$

$$\{ \overrightarrow{\text{plur}} \cdot E(\text{puppy}) = -2.95$$

$$E(\text{queen}) \approx E(\text{king}) + E(\text{woman}) - E(\text{man})$$



quill



Harry Potter was a highly unusual boy in many ways. For one thing, he hated the summer holidays more than any other time of year. For another, he really wanted to do his homework but was forced to do it in secret, in the dead of night. And he also happened to be a wizard.

It was nearly midnight, and he was lying on his stomach in bed, the blankets drawn right over his head like a tent, a flashlight in one hand and a large leather-bound book (*A History of Magic* by Bathilda Bagshot) propped open against the pillow. Harry moved the tip of his eagle-feather quill down the page, frowning as he looked for something that would help him write his essay, "Witch Burning in the Fourteenth Century Was Completely Pointless discuss."

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Harry put his quill between his teeth and reached underneath his pillow for his ink bottle and a roll of parchment. Slowly and very carefully he unscrewed the ink bottle, dipped his quill into it, and began to write, pausing every now and then to listen, because if any of the Dursleys heard the scratching of his quill on their way to the bathroom, he'd probably find himself locked in the cupboard under the stairs for the rest of the summer.

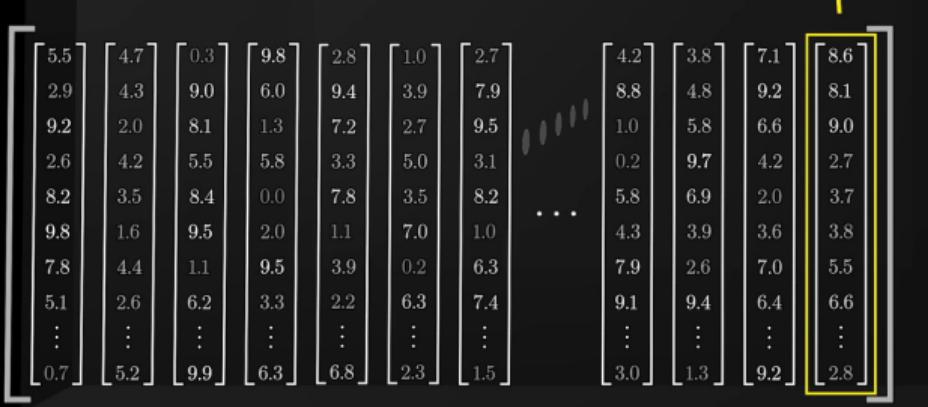
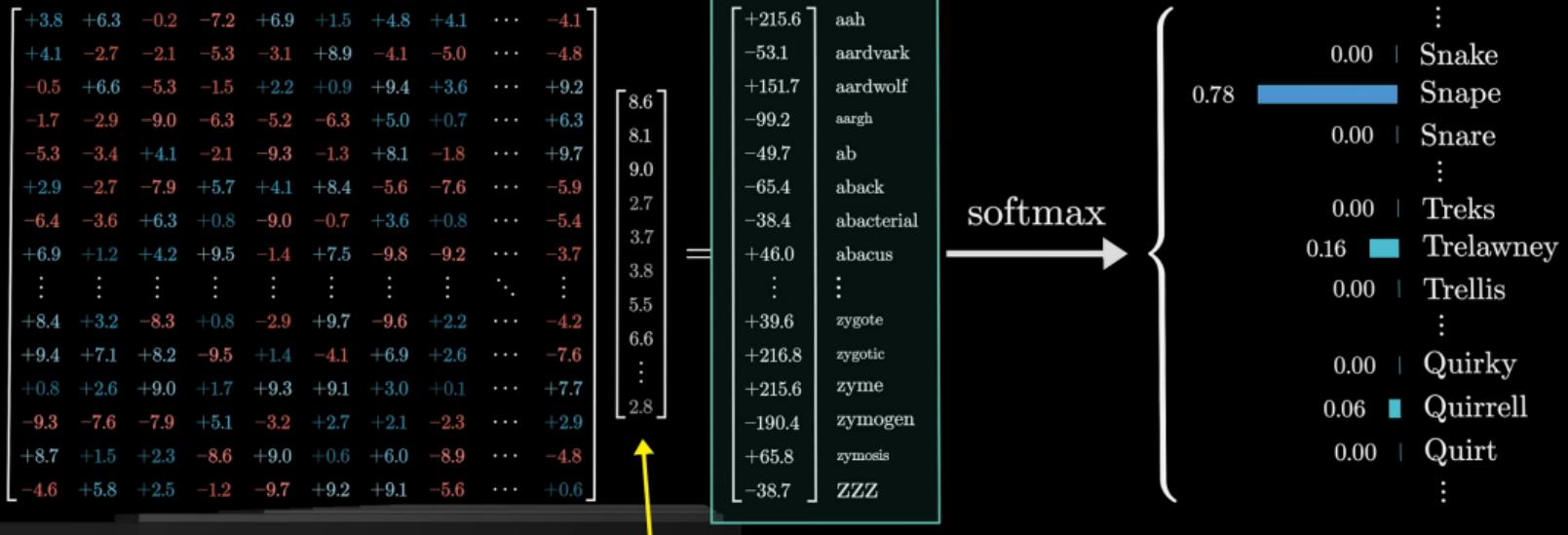
The Dursley family of number four, Privet Drive, was the reason that Harry never enjoyed his summer holidays. Uncle Vernon, Aunt Petunia, and their son, Dudley, were Harry's only living relatives. They were Muggles, and they had a very medieval attitude toward magic. Harry's dead parents, who had been a witch and wizard themselves, were never mentioned under the Dursleys' roof For years, Aunt Petunia and Uncle Vernon had hoped that if they kept Harry as mowntrodded as possible, they would be able to squash the magic out of him. To their fury, they had been unsuccessful. These days they lived in terror of anyone finding out that Harry had spent most of the last two years at Hogwarts School of Witchcraft and Wizardry. The most they could do, however, was to lock away Harry's spellbooks, wand, cauldron, and broomstick at the start of the summer break, and forbid him to talk to the neighbors.

This separation from his spellbooks had been a real problem for Harry, because his teachers at Hogwarts had given him a lot of holiday work. One of the essays, a particularly nasty one about shrinking potions, was for Harry's least favorite teacher, Professor

Harry Potter was a highly unusual boy. He lived in a small town called Privet Drive. ...

5.5	4.7	0.3	9.8	2.8	1.0	2.7	4.2	3.8	7.1	8.6
2.9	4.3	9.0	6.0	9.4	3.9	7.9	8.8	4.8	9.2	8.1
9.2	2.0	8.1	1.3	7.2	2.7	9.5	1.0	5.8	6.6	9.0
2.6	4.2	5.5	5.8	3.3	5.0	3.1	0.2	9.7	4.2	2.7
8.2	3.5	8.4	0.0	7.8	3.5	8.2	5.8	6.9	2.0	3.7
9.8	1.6	9.5	2.0	1.1	7.0	1.0	4.3	3.9	3.6	3.8
7.8	4.4	1.1	9.5	3.9	0.2	6.3	7.9	2.6	7.0	5.5
5.1	2.6	6.2	3.3	2.2	6.3	7.4	9.1	9.4	6.4	6.6
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
0.7	5.2	9.9	6.3	6.8	2.3	1.5	3.0	1.3	9.2	2.8

Context size = 2,048



Total weights: 175,181,291,520
Organized into 27,938 matrices



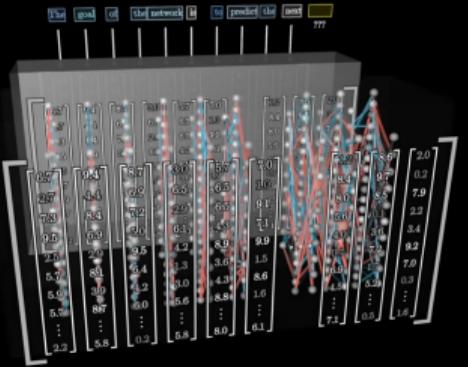
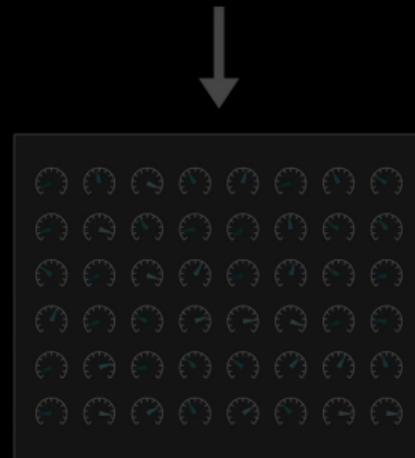
Embedding	$d_{\text{embed}} * n_{\text{vocab}} = 12,288 * 50,257 = 617,558,016$
Key	
Query	
Value	
Output	
Up-projection	
Down-projection	
Unembedding	$n_{\text{vocab}} * d_{\text{embed}} = 50,257 * 12,288 = 617,558,016$

1,235,116,032

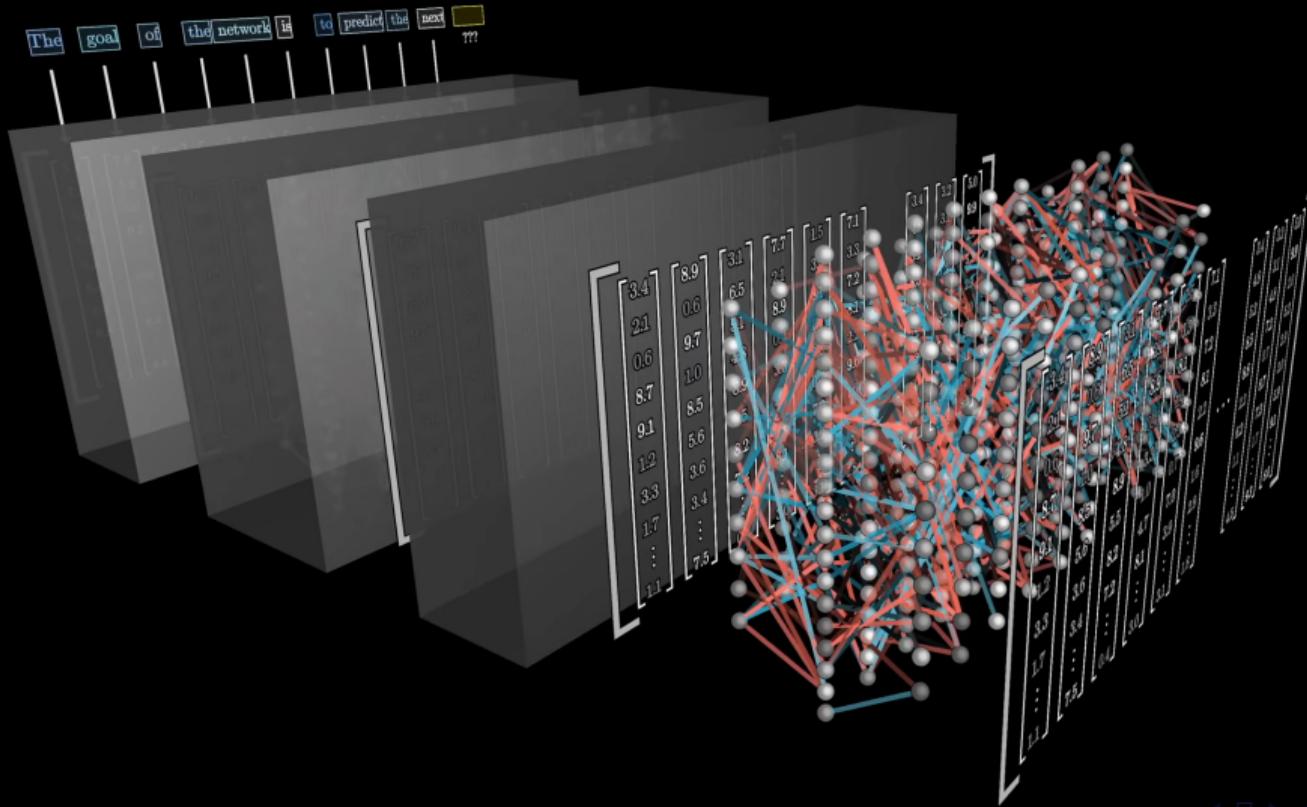
Generative Pre-trained Transformer

computer science is to practice coding regularly. Start by working on simple exercises and gradually move on to more complex projects. 4. Participate in coding challenges and competitions: Coding challenges and competitions provide a great opportunity to put your skills to the test and learn from others

drew back from the window, and the Doctor looked for explanation in his friend's ashy face. "They are," Mr. Lorry whispered the words, glancing fearfully round at the locked room, "murdering the prisoners. If you are sure of what you say; if you really have the power you think you



Transformer



Attention Is All You Need

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Abstract

The dominant sequence transduction models are based on complex recurrent or convolutional neural networks that include an encoder and a decoder. The best performing models also connect the encoder and decoder through an attention mechanism. We propose a new simple network architecture, the Transformer, based solely on attention mechanisms, dispensing with recurrence and convolutions entirely. Experiments on two machine translation tasks show these models to be superior in quality while being more parallelizable and requiring significantly less time to train. Our model achieves 28.4 BLEU on the WMT 2014 English-to-German translation task, improving over the existing best results, including ensembles, by over 2 BLEU. On the WMT 2014 English-to-French translation task, our model establishes a new single-model state-of-the-art BLEU score of 41.8 after training for 3.5 days on eight GPUs, a small fraction of the training costs of the best models from the literature. We show that the Transformer generalizes well to

A Convenient Lie

Let's pretend that tokens are always simply words

The Truth

This process (known fancifully as tokenization) frequently subdivides words



American shrew  mole

6.02×10^{23}

One  mole of carbon dioxide



Take a biopsy of the  mole

American shrew mole

↓	↓	↓
6.0	0.4	5.8
2.2	5.7	9.9
3.9	5.0	2.5
7.7	1.8	3.7
6.1	9.7	9.1
:	:	:
6.3	5.4	2.1

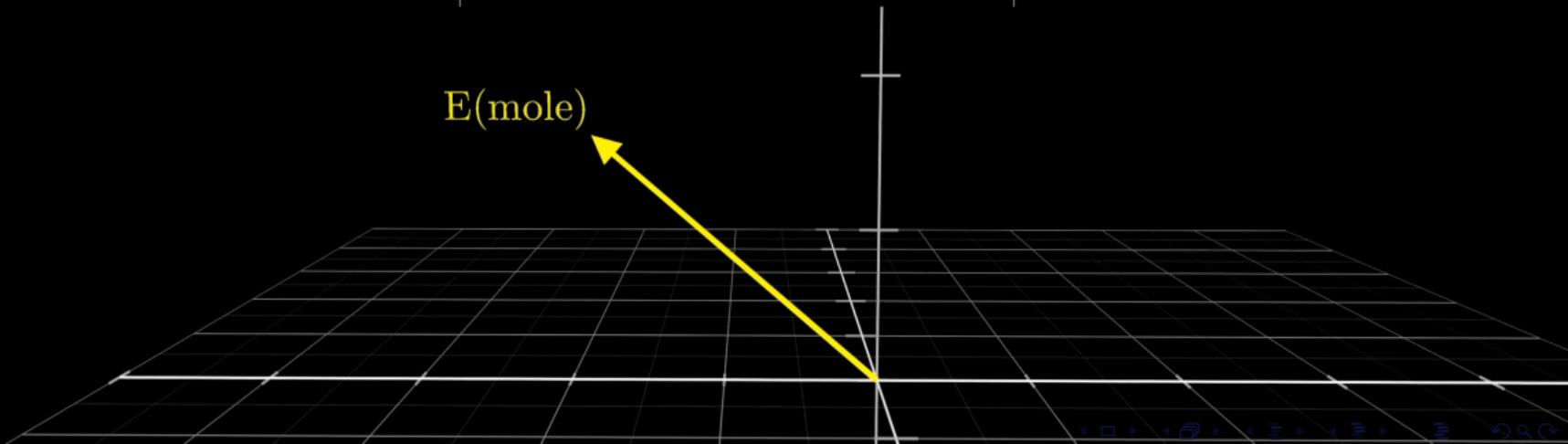
One mole of carbon dioxide

↓	↓	↓	↓	↓
5.2	5.8	5.8	7.6	9.9
7.8	9.9	7.0	4.5	1.8
2.5	2.5	4.0	5.7	6.1
5.9	3.7	0.1	8.1	9.8
9.8	9.1	4.3	5.6	9.1
:	:	:	:	:
2.7	2.1	4.5	4.8	0.4

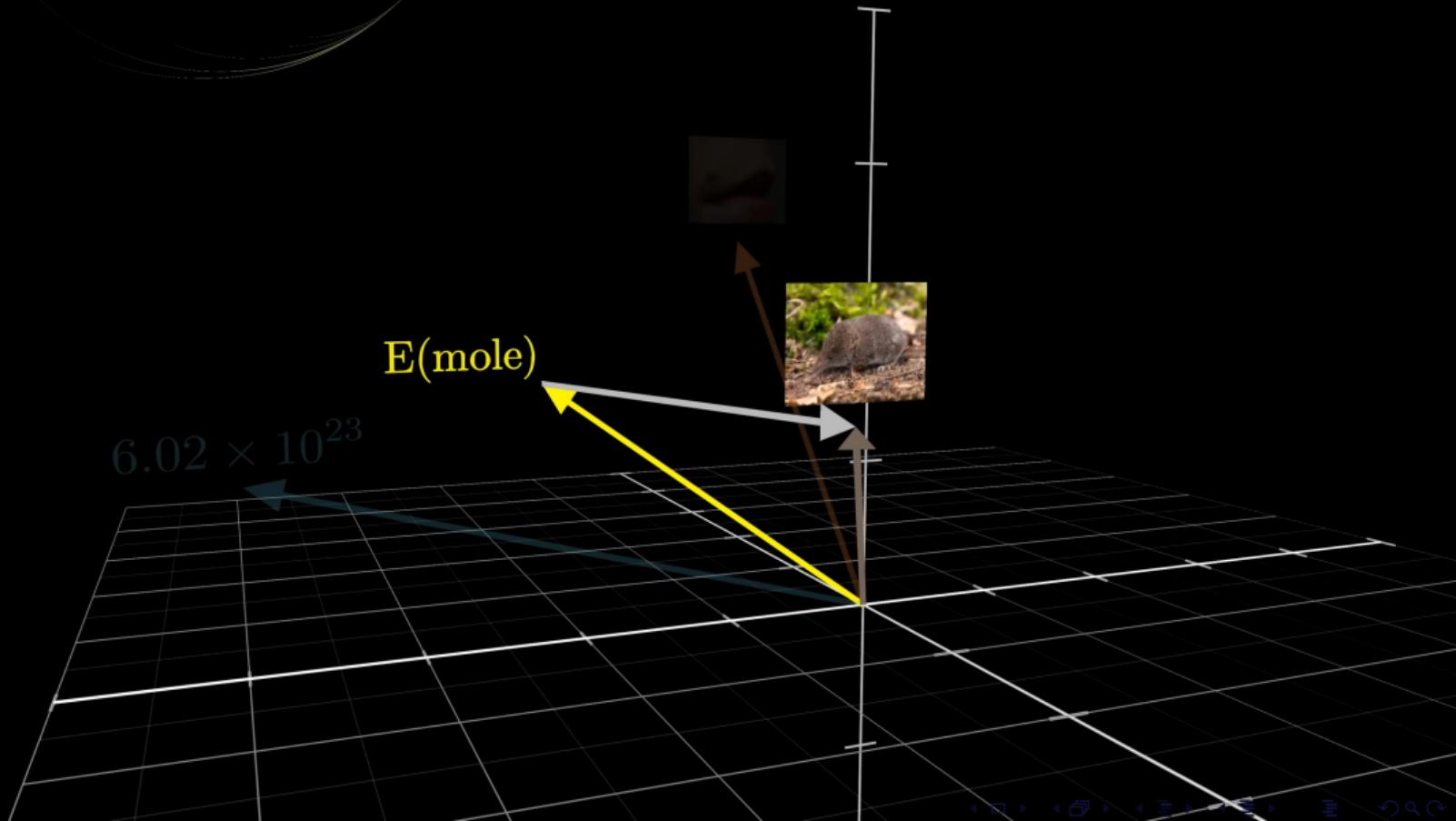
Take a biopsy of the mole

↓	↓	↓	↓	↓	↓
4.9	3.5	1.7	5.8	2.3	5.8
2.1	9.7	8.7	7.0	4.9	9.9
4.7	3.6	3.4	4.0	6.4	2.5
9.6	8.3	2.7	0.1	3.2	3.7
8.0	0.8	4.7	4.3	4.4	9.1
:	:	:	:	:	:
2.2	8.9	2.3	4.5	6.5	2.1

E(mole)



American shrew mole



One mole of carbon dioxide

$E(\text{mole})$

6.02×10^{23}



Short distance

One mole of carbon dioxide



Long distance

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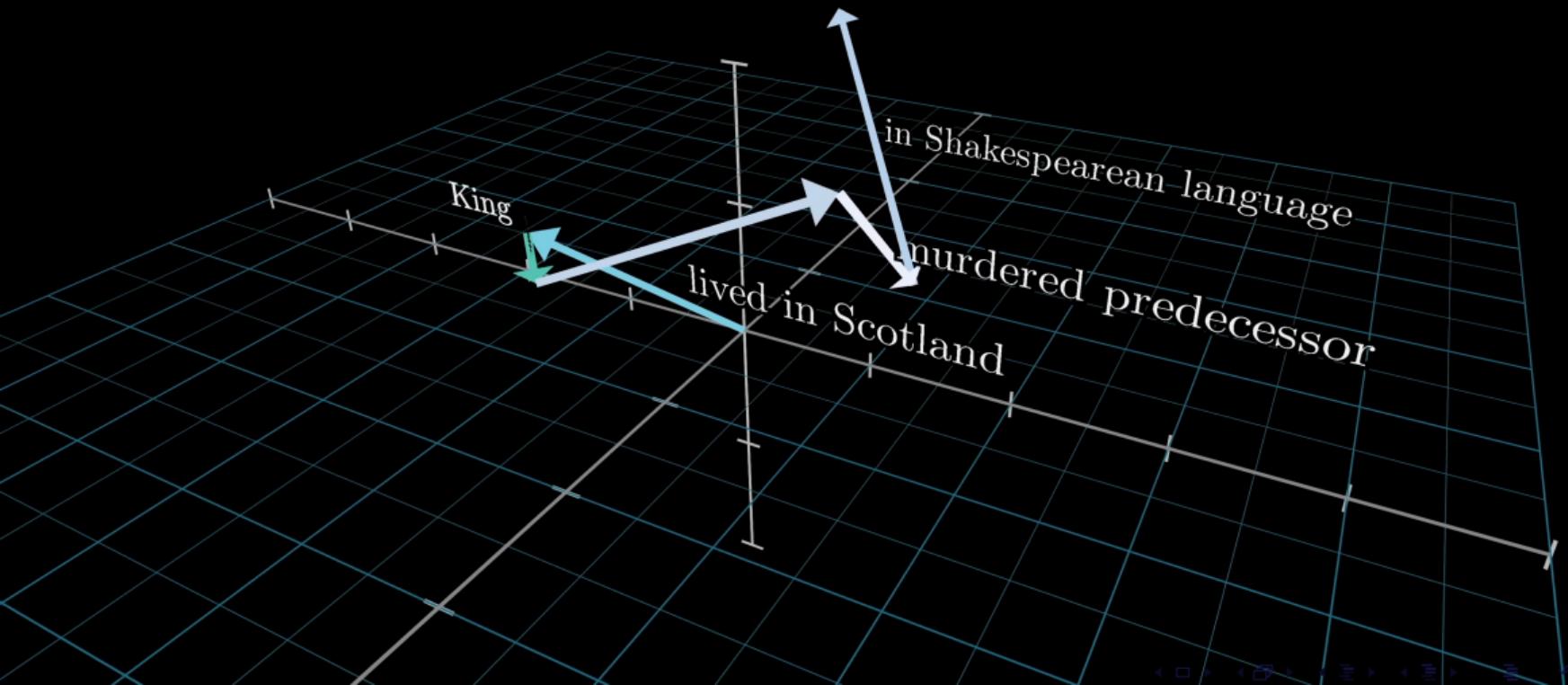
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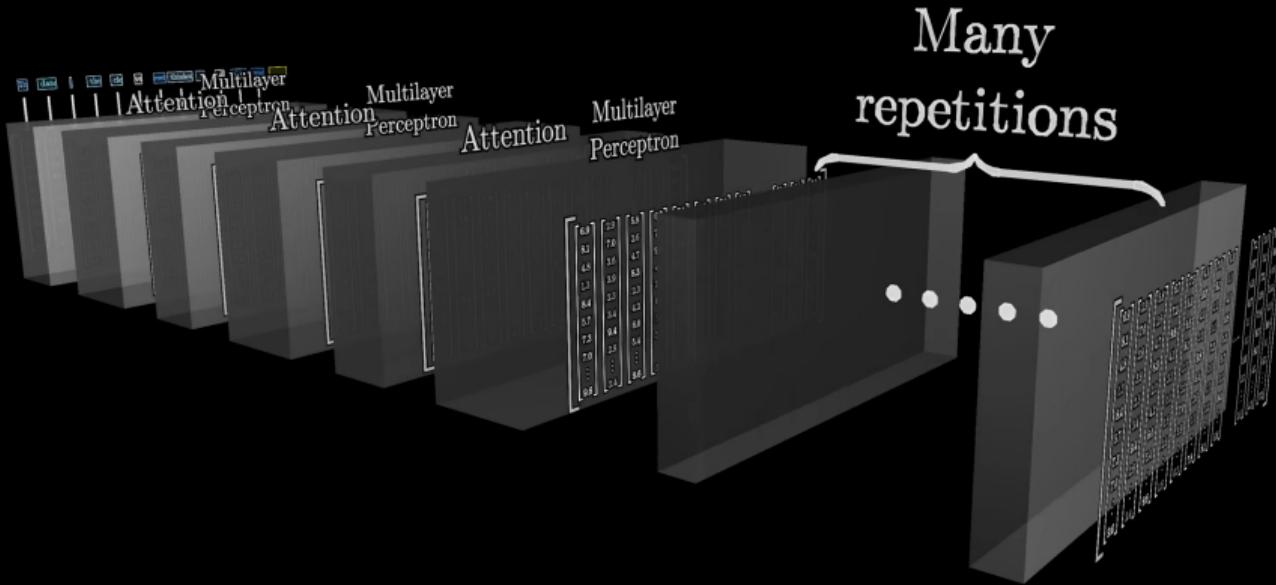
Harry put his quill between his teeth and reached underneath his pillow for his ink bottle and a roll of parchment. Slowly and very carefully he unscrewed the ink bottle, dipped his quill into it, and began to write, pausing every now and then to listen, because if any of the Dursleys heard the scratching of his quill on their way to the bathroom, he'd probably find himself locked in the cupboard under the stairs for the rest of the summer.

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The King doth wake tonight and takes his rouse ...





0.9	0.7	0.6	0.5	0.4	0.3	0.2	0.1	0.0
0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9
0.5	0.4	0.3	0.2	0.1	0.0	0.9	0.8	0.7
0.3	0.2	0.1	0.0	0.9	0.8	0.7	0.6	0.5
0.7	0.6	0.5	0.4	0.3	0.2	0.1	0.0	0.9
0.2	0.1	0.0	0.9	0.8	0.7	0.6	0.5	0.4
0.8	0.7	0.6	0.5	0.4	0.3	0.2	0.1	0.0
0.6	0.5	0.4	0.3	0.2	0.1	0.0	0.9	0.8
0.4	0.3	0.2	0.1	0.0	0.9	0.8	0.7	0.6
0.9	0.8	0.7	0.6	0.5	0.4	0.3	0.2	0.1



- the ■ 8.82%
probably ■ 4.37%
John ■ 4.04%
Sir ■ 3.66%
Albert ■ 3.63%
Ber ■ 3.31%
a ■ 2.90%
Isaac ■ 2.01%
undoubtedly ■ 1.58%
arguably ■ 1.33%
Im ■ 1.16%
Einstein ■ 1.13%
Ludwig ■ 1.04%
::

The guests were Adam, Jonathan's business partner; Bella, Jonathan's estranged sister; Claire, the caretaker of the mansion; and Derek, an old college friend. Sarah began her investigation, noting the muddy footprints that led to Jonathan's room, the absence of Adam's raincoat, Bella's uncontrollable sobbing, Claire's nervous fidgeting, and Derek's calm demeanor.

Adam had motive, given the recent disputes over their business, but his alibi was solid, having been in the crowded living room when the murder occurred. Bella, despite her estrangement, had been seen on the opposite side of the mansion. Claire had access to all rooms but lacked any motive. Derek, however, had been mysteriously absent; his whereabouts unaccounted for.

•
•
•

"Each of you has what on the surface appears an unassimilable alibi," she said, as the storm cleared and the crimson rays of dusk poured through the window where everyone was gathered. "But only one of you could have known about the loose screw on the window, while also having known where the second key was hidden. I am left inescapably shown, to the conclusion that therefore, the murderer was [redacted] ??"



[+6.1]
+3.7
+2.6
+3.4
+2.1
+2.8
+2.4
+0.7
+5.1
-2.6
⋮
-0.5

a **fluffy** **blue** creature roamed the **verdant** forest





a	fluffy	blue	creature	roamed	the	verdant	forest
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Position

1

2

3

4

5

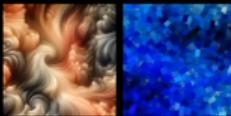
6

7

8

12,288

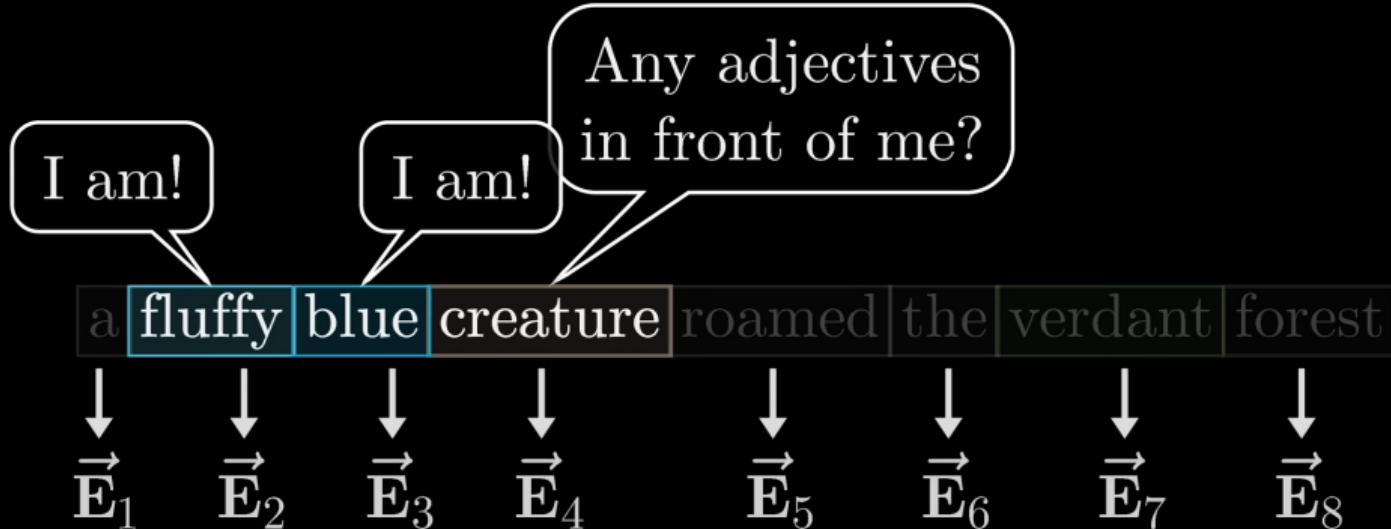
$$\left\{ \begin{array}{c} \begin{bmatrix} 6.0 \\ 0.2 \\ 3.0 \\ 6.5 \\ 2.9 \\ 6.1 \\ 4.2 \\ 1.3 \\ \vdots \\ 3.0 \end{bmatrix}, \begin{bmatrix} 5.6 \\ 5.8 \\ 5.7 \\ 6.5 \\ 6.5 \\ 4.3 \\ 8.9 \\ 3.6 \\ \vdots \\ 4.3 \end{bmatrix}, \begin{bmatrix} 8.8 \\ 8.0 \\ 7.0 \\ 1.0 \\ 9.1 \\ 7.1 \\ 9.9 \\ 1.5 \\ \vdots \\ 8.6 \end{bmatrix}, \begin{bmatrix} 1.6 \\ 6.1 \\ 1.2 \\ 8.4 \\ 8.0 \\ 5.6 \\ 4.0 \\ 0.7 \\ \vdots \\ 6.9 \end{bmatrix}, \begin{bmatrix} 4.5 \\ 7.1 \\ 8.6 \\ 9.7 \\ 8.5 \\ 0.1 \\ 3.6 \\ 7.2 \\ \vdots \\ 1.7 \end{bmatrix}, \begin{bmatrix} 5.2 \\ 0.5 \\ 2.0 \\ 0.2 \\ 7.9 \\ 2.2 \\ 3.4 \\ 9.2 \\ \vdots \\ 7.0 \end{bmatrix}, \begin{bmatrix} 0.3 \\ 1.6 \\ 6.2 \\ 5.7 \\ 2.4 \\ 9.2 \\ 6.1 \\ 5.3 \\ \vdots \\ 5.8 \end{bmatrix}, \begin{bmatrix} 7.2 \\ 3.1 \\ 3.9 \\ 2.1 \\ 1.8 \\ 9.3 \\ 7.3 \\ 4.9 \\ \vdots \\ 2.3 \end{bmatrix} \end{array} \right.$$



a fluffy blue creature roamed the verdant forest

$$\vec{\mathbf{E}}_1' \quad \vec{\mathbf{E}}_2' \quad \vec{\mathbf{E}}_3' \quad \vec{\mathbf{E}}_4' \quad \vec{\mathbf{E}}_5' \quad \vec{\mathbf{E}}_6' \quad \vec{\mathbf{E}}_7' \quad \vec{\mathbf{E}}_8'$$





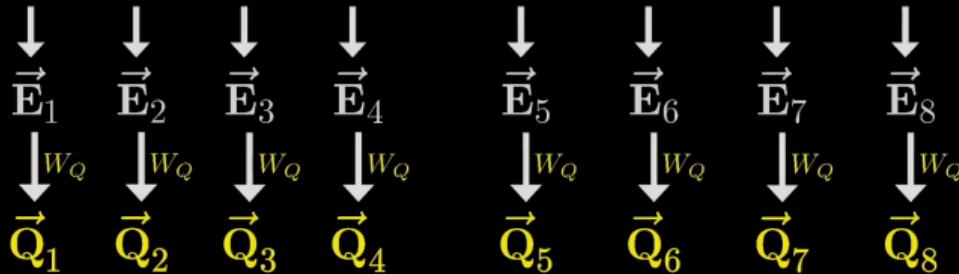
a	fluffy	blue	creature	roamed	the	verdant	forest
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$$\begin{array}{cccccccc}
 \downarrow & \downarrow \\
 \vec{\mathbf{E}}_1 & \vec{\mathbf{E}}_2 & \vec{\mathbf{E}}_3 & \vec{\mathbf{E}}_4 & \vec{\mathbf{E}}_5 & \vec{\mathbf{E}}_6 & \vec{\mathbf{E}}_7 & \vec{\mathbf{E}}_8 \\
 \downarrow W_Q & \downarrow W_Q \\
 \vec{\mathbf{Q}}_1 & \vec{\mathbf{Q}}_2 & \vec{\mathbf{Q}}_3 & \vec{\mathbf{Q}}_4 & \vec{\mathbf{Q}}_5 & \vec{\mathbf{Q}}_6 & \vec{\mathbf{Q}}_7 & \vec{\mathbf{Q}}_8
 \end{array}$$

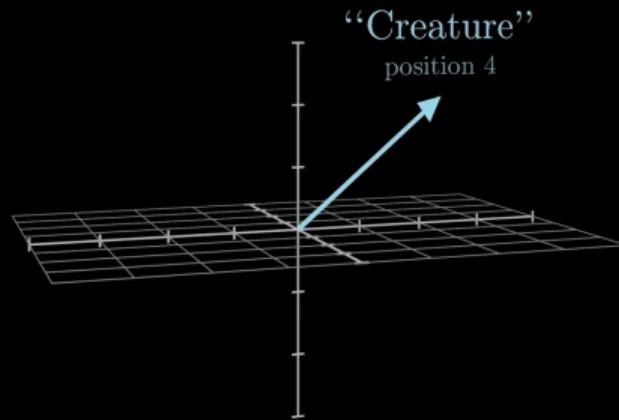
Any adjectives
in front of me?

$$\underbrace{W_Q}_{\begin{bmatrix} +7.5 & -3.2 & +9.1 & -5.3 & +8.9 & +8.7 & +5.9 & +2.6 & +7.4 & -4.1 & \cdots & +2.3 \\ -9.6 & -3.0 & -7.0 & +9.5 & -0.4 & -0.1 & +2.8 & -2.6 & -7.2 & +6.4 & \cdots & +0.2 \\ -5.5 & -8.0 & +7.2 & +9.4 & +9.1 & +8.0 & +5.4 & -3.3 & -8.3 & -1.8 & \cdots & -7.3 \\ -8.8 & +4.5 & -9.7 & +5.4 & -7.0 & -8.3 & -8.1 & +3.4 & +5.0 & -1.6 & \cdots & +7.1 \\ +4.5 & -4.5 & -7.3 & -8.8 & -3.9 & -4.7 & -0.9 & +3.6 & +3.9 & -4.3 & \cdots & -6.3 \\ \vdots & \ddots & \cdots & \vdots \\ -9.0 & +5.9 & -8.4 & +0.4 & -3.8 & +1.5 & +9.1 & +2.9 & -9.2 & -1.4 & \cdots & +0.7 \end{bmatrix}} \vec{\mathbf{E}}_i = \vec{\mathbf{Q}}_i$$

a|fluffy|blue|creature|roamed|the|verdant|forest

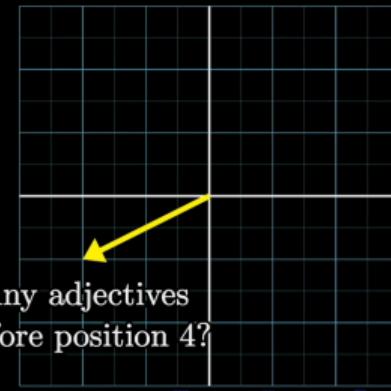


Embedding space
12,288-dimensional



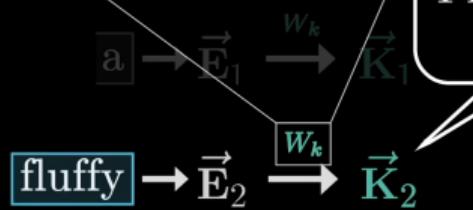
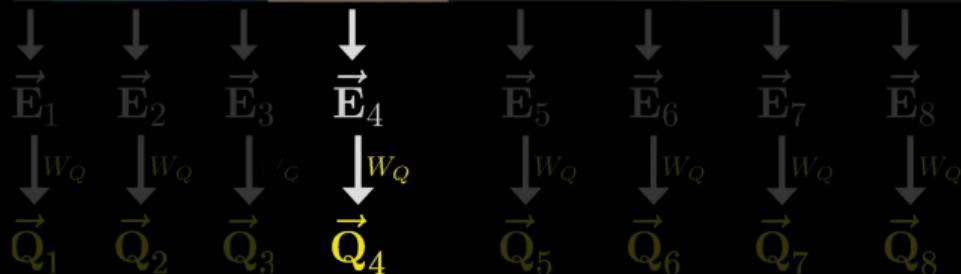
$$W_Q \rightarrow$$

Query/Key space
128-dimensional

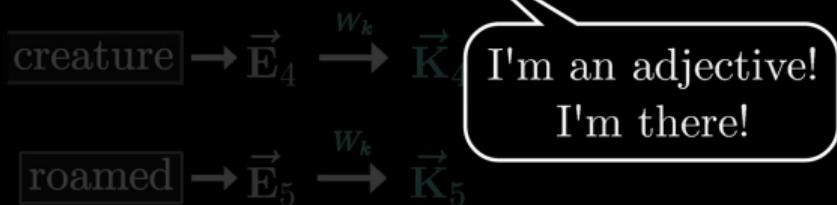
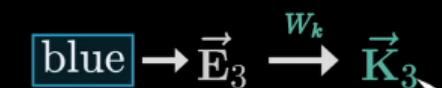


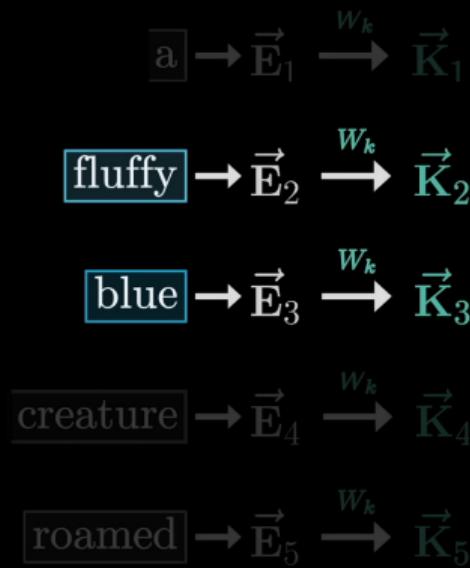
a|fluffy|blue|creature|roamed|the|verdant|forest

-7.6	-4.5	-1.9	-2.0	+3.4	-3.1	+4.2	+2.8	-2.0	-1.4	...	+2.1
-8.5	+6.4	+3.0	+4.5	+0.7	-7.7	-1.9	-1.9	-3.5	-9.3	...	+4.7
-7.7	+2.1	+4.0	+2.7	+9.1	-7.9	+7.3	-9.3	+0.7	-1.9	...	+0.5
-2.7	-6.1	-9.5	+0.4	+6.8	-2.5	-5.5	-8.3	-8.2	-5.5	...	-7.9
-4.7	-8.6	-8.6	+7.1	-6.7	+1.2	+5.4	-0.9	-6.9	-5.9	...	-1.3
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	...	⋮
+0.6	-3.0	+5.6	+5.0	+8.5	-9.3	+7.8	-2.1	+7.5	+3.8	...	+9.6

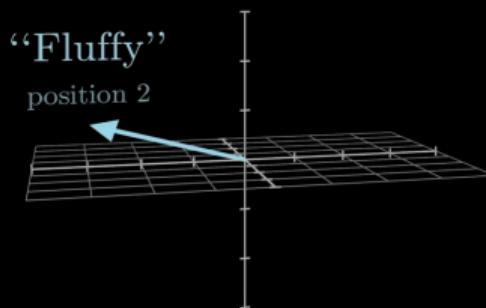


I'm an adjective!
I'm there!





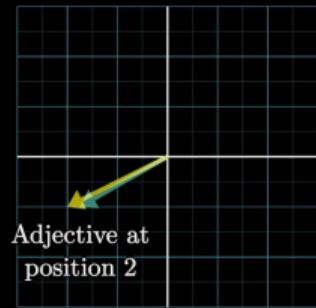
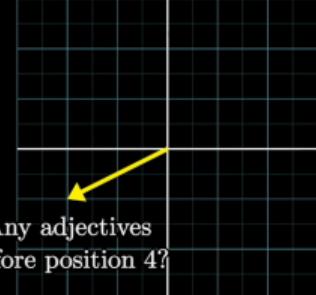
Embedding space 12,288-dimensional



Query/Key space 128-dimensional

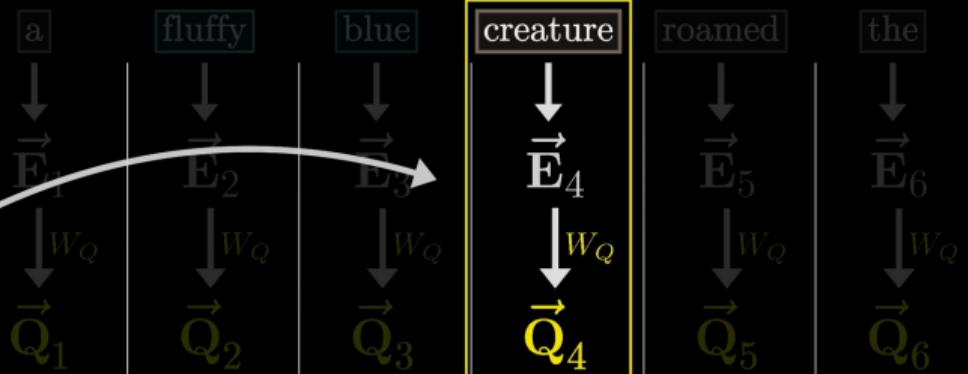


Adjective at
position 2



a	fluffy	blue	creature	roamed	the	verdant	forest	
\downarrow \vec{E}_1 $\downarrow W_Q$ \vec{Q}_1	\downarrow \vec{E}_2 $\downarrow W_Q$ \vec{Q}_2	\downarrow \vec{E}_3 $\downarrow W_Q$ \vec{Q}_3	\downarrow \vec{E}_4 $\downarrow W_Q$ \vec{Q}_4	\downarrow \vec{E}_5 $\downarrow W_Q$ \vec{Q}_5	\downarrow \vec{E}_6 $\downarrow W_Q$ \vec{Q}_6	\downarrow \vec{E}_7 $\downarrow W_Q$ \vec{Q}_7	\downarrow \vec{E}_8 $\downarrow W_Q$ \vec{Q}_8	
$[a] \rightarrow \vec{E}_1 \xrightarrow{W_k} \vec{K}_1$	$\vec{K}_1 \cdot \vec{Q}_1$	$\vec{K}_1 \cdot \vec{Q}_2$	$\vec{K}_1 \cdot \vec{Q}_3$	$\vec{K}_1 \cdot \vec{Q}_4$	$\vec{K}_1 \cdot \vec{Q}_5$	$\vec{K}_1 \cdot \vec{Q}_6$	$\vec{K}_1 \cdot \vec{Q}_7$	$\vec{K}_1 \cdot \vec{Q}_8$
$[\text{fluffy}] \rightarrow \vec{E}_2 \xrightarrow{W_k} \vec{K}_2$	$\vec{K}_2 \cdot \vec{Q}_1$	$\vec{K}_2 \cdot \vec{Q}_2$	$\vec{K}_2 \cdot \vec{Q}_3$	$\vec{K}_2 \cdot \vec{Q}_4$	$\vec{K}_2 \cdot \vec{Q}_5$	$\vec{K}_2 \cdot \vec{Q}_6$	$\vec{K}_2 \cdot \vec{Q}_7$	$\vec{K}_2 \cdot \vec{Q}_8$
$[\text{blue}] \rightarrow \vec{E}_3 \xrightarrow{W_k} \vec{K}_3$	$\vec{K}_3 \cdot \vec{Q}_1$	$\vec{K}_3 \cdot \vec{Q}_2$	$\vec{K}_3 \cdot \vec{Q}_3$	$\vec{K}_3 \cdot \vec{Q}_4$	$\vec{K}_3 \cdot \vec{Q}_5$	$\vec{K}_3 \cdot \vec{Q}_6$	$\vec{K}_3 \cdot \vec{Q}_7$	$\vec{K}_3 \cdot \vec{Q}_8$
$[\text{creature}] \rightarrow \vec{E}_4 \xrightarrow{W_k} \vec{K}_4$	$\vec{K}_4 \cdot \vec{Q}_1$	$\vec{K}_4 \cdot \vec{Q}_2$	$\vec{K}_4 \cdot \vec{Q}_3$	$\vec{K}_4 \cdot \vec{Q}_4$	$\vec{K}_4 \cdot \vec{Q}_5$	$\vec{K}_4 \cdot \vec{Q}_6$	$\vec{K}_4 \cdot \vec{Q}_7$	$\vec{K}_4 \cdot \vec{Q}_8$
$[\text{roamed}] \rightarrow \vec{E}_5 \xrightarrow{W_k} \vec{K}_5$	$\vec{K}_5 \cdot \vec{Q}_1$	$\vec{K}_5 \cdot \vec{Q}_2$	$\vec{K}_5 \cdot \vec{Q}_3$	$\vec{K}_5 \cdot \vec{Q}_4$	$\vec{K}_5 \cdot \vec{Q}_5$	$\vec{K}_5 \cdot \vec{Q}_6$	$\vec{K}_5 \cdot \vec{Q}_7$	$\vec{K}_5 \cdot \vec{Q}_8$
$[\text{the}] \rightarrow \vec{E}_6 \xrightarrow{W_k} \vec{K}_6$	$\vec{K}_6 \cdot \vec{Q}_1$	$\vec{K}_6 \cdot \vec{Q}_2$	$\vec{K}_6 \cdot \vec{Q}_3$	$\vec{K}_6 \cdot \vec{Q}_4$	$\vec{K}_6 \cdot \vec{Q}_5$	$\vec{K}_6 \cdot \vec{Q}_6$	$\vec{K}_6 \cdot \vec{Q}_7$	$\vec{K}_6 \cdot \vec{Q}_8$
$[\text{verdant}] \rightarrow \vec{E}_7 \xrightarrow{W_k} \vec{K}_7$	$\vec{K}_7 \cdot \vec{Q}_1$	$\vec{K}_7 \cdot \vec{Q}_2$	$\vec{K}_7 \cdot \vec{Q}_3$	$\vec{K}_7 \cdot \vec{Q}_4$	$\vec{K}_7 \cdot \vec{Q}_5$	$\vec{K}_7 \cdot \vec{Q}_6$	$\vec{K}_7 \cdot \vec{Q}_7$	$\vec{K}_7 \cdot \vec{Q}_8$
$[\text{forest}] \rightarrow \vec{E}_8 \xrightarrow{W_k} \vec{K}_8$	$\vec{K}_8 \cdot \vec{Q}_1$	$\vec{K}_8 \cdot \vec{Q}_2$	$\vec{K}_8 \cdot \vec{Q}_3$	$\vec{K}_8 \cdot \vec{Q}_4$	$\vec{K}_8 \cdot \vec{Q}_5$	$\vec{K}_8 \cdot \vec{Q}_6$	$\vec{K}_8 \cdot \vec{Q}_7$	$\vec{K}_8 \cdot \vec{Q}_8$

“Attend to”



$$[a] \rightarrow \vec{E}_1 \xrightarrow{W_k} \vec{K}_1$$

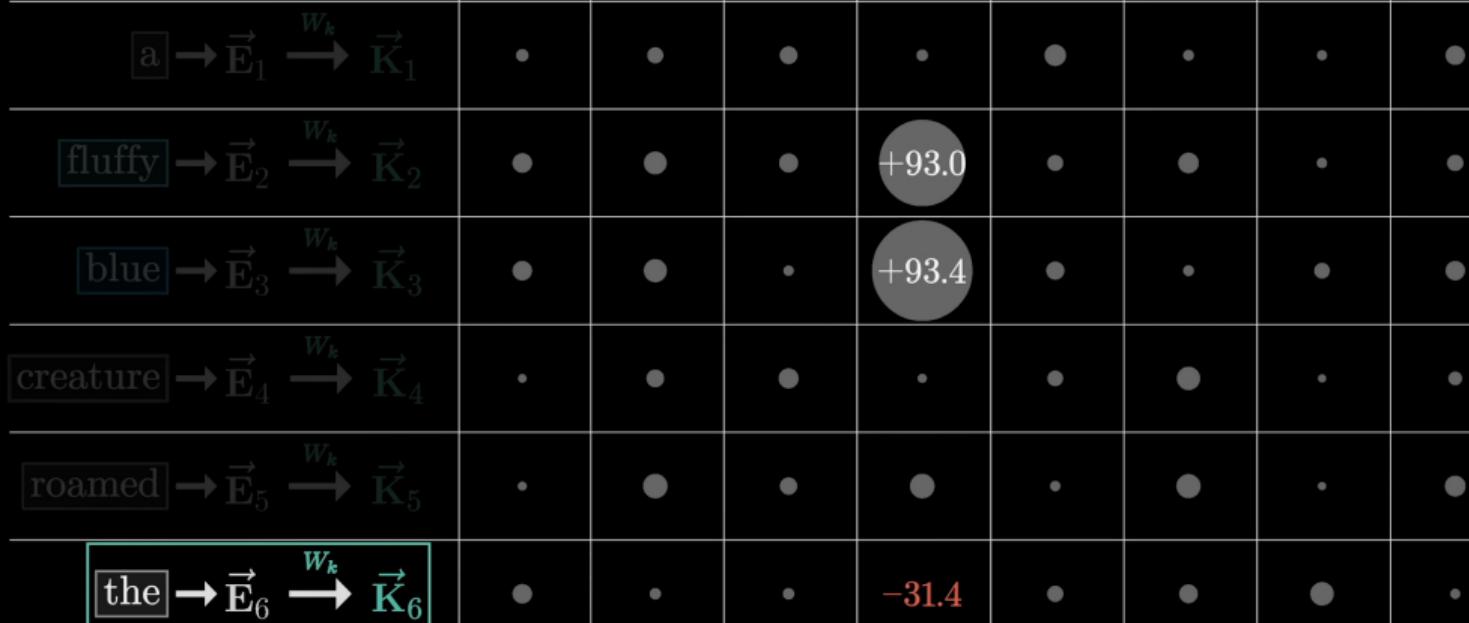
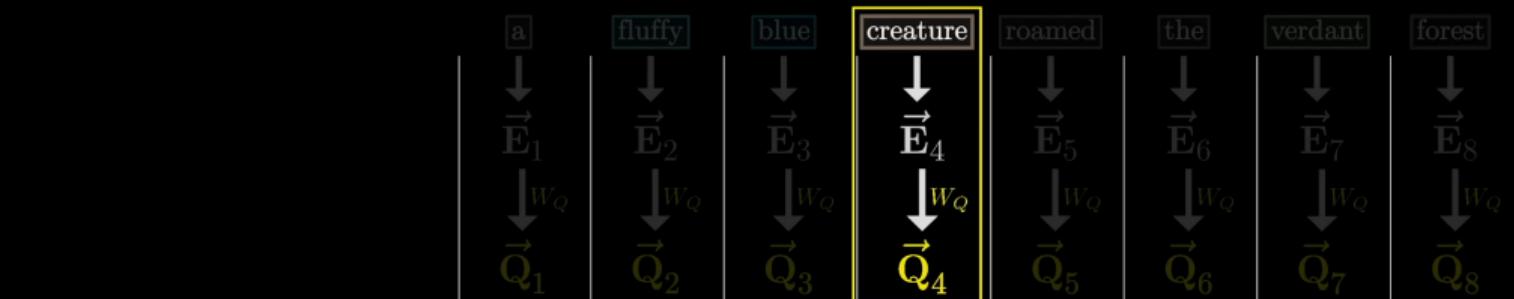
$$\text{fluffy} \rightarrow \vec{E}_2 \xrightarrow{W_k} \vec{K}_2$$

$$\text{blue} \rightarrow \vec{E}_3 \xrightarrow{W_k} \vec{K}_3$$

$$\text{creature} \rightarrow \vec{E}_4 \xrightarrow{W_k} \vec{K}_4$$

+93.0

+93.4



	\vec{E}_1	\vec{E}_2	\vec{E}_3	\vec{E}_4	\vec{E}_5	\vec{E}_6	\vec{E}_7	\vec{E}_8	
a	\vec{Q}_1	\vec{Q}_2	\vec{Q}_3	\vec{Q}_4	\vec{Q}_5	\vec{Q}_6	\vec{Q}_7	\vec{Q}_8	
fluffy									
blue									
creature									
roamed									
the									
verdant									
forest									
\vec{K}_1									
\vec{K}_2									
\vec{K}_3									
\vec{K}_4									
\vec{K}_5									
\vec{K}_6									
\vec{K}_7									
\vec{K}_8									

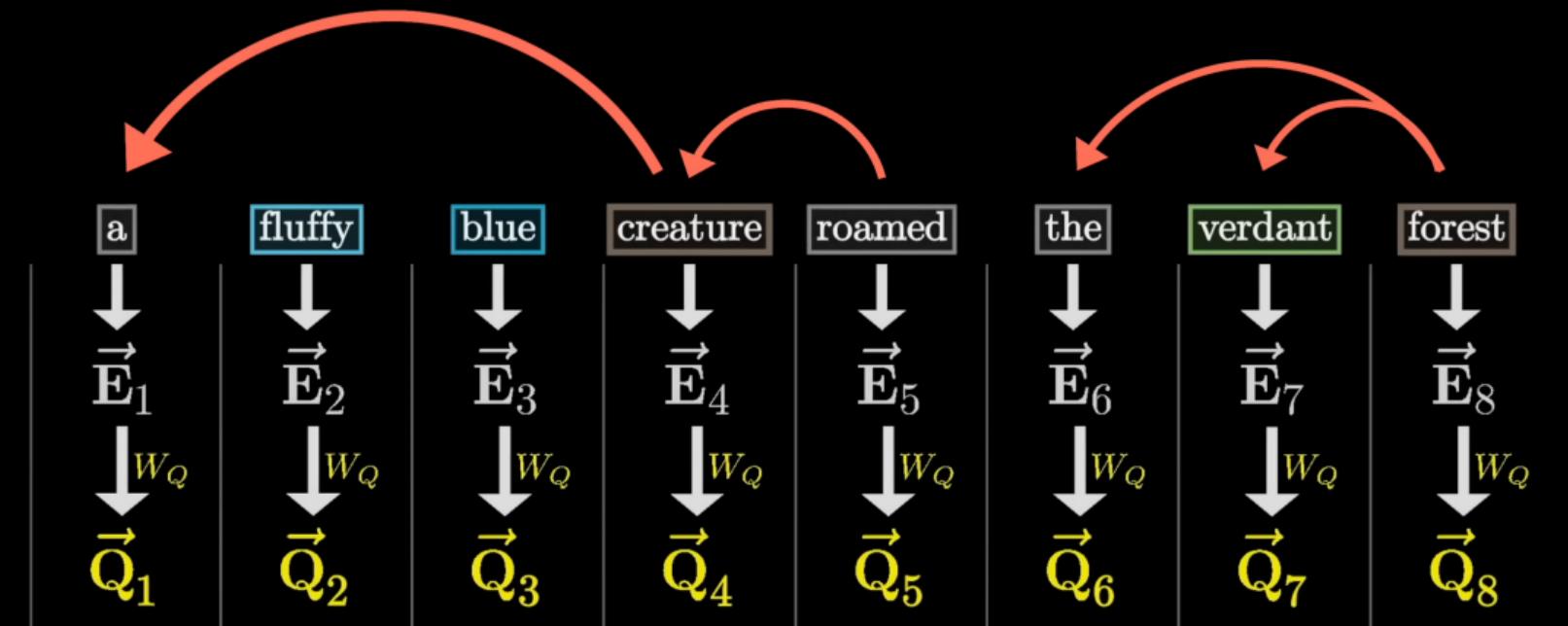
$\vec{a} \rightarrow \vec{E}_1 \xrightarrow{w_k} \vec{K}_1$	+0.7	-83.7	-24.7	-27.8	-5.2	-89.3	45.2	-36.1	0.00
$\text{fluffy} \rightarrow \vec{E}_2 \xrightarrow{w_k} \vec{K}_2$	-73.4	+2.9	-5.4	+93.0	-48.2	-87.1	-97.5	-7.8	0.42
$\text{blue} \rightarrow \vec{E}_3 \xrightarrow{w_k} \vec{K}_3$	-53.4	-5.7	+1.8	+93.4	-55.6	-56.0	-96.1	-0.1	0.58
$\text{creature} \rightarrow \vec{E}_4 \xrightarrow{w_k} \vec{K}_4$	-21.5	-29.7	-56.1	+4.9	-32.4	-92.3	95	-28.1	0.00
$\text{roamed} \rightarrow \vec{E}_5 \xrightarrow{w_k} \vec{K}_5$	-20.1	-40.9	-87.8	-55.4	-0.6	-1.6	-1.6	-98.8	0.00
$\text{the} \rightarrow \vec{E}_6 \xrightarrow{w_k} \vec{K}_6$	-87.9	-33.3	-22.0	-31.4	-5.5	-0.6	-1.6	-98.8	0.00
$\text{verdant} \rightarrow \vec{E}_7 \xrightarrow{w_k} \vec{K}_7$	-41.2	-55.5	-42.3	-59.8	-79.0	-97.9	-93.7	-43.8	0.00
$\text{forest} \rightarrow \vec{E}_8 \xrightarrow{w_k} \vec{K}_8$	-58.9	-75.5	-91.1	-90.6	-75.6	-89.0	-70.8	-47	0.00

softmax



$$\text{Attention}(Q, K, V) = \text{softmax}\left(\frac{QK^T}{\sqrt{d_k}}\right)V$$

	Q_1	Q_2	Q_3	Q_4	Q_5	\dots	Q_n
K_1	$\frac{Q_1 \cdot K_1}{\sqrt{d_k}}$	$\frac{Q_2 \cdot K_1}{\sqrt{d_k}}$	$\frac{Q_3 \cdot K_1}{\sqrt{d_k}}$	$\frac{Q_4 \cdot K_1}{\sqrt{d_k}}$	$\frac{Q_5 \cdot K_1}{\sqrt{d_k}}$	\dots	$\frac{Q_n \cdot K_1}{\sqrt{d_k}}$
K_2	$\frac{Q_1 \cdot K_2}{\sqrt{d_k}}$	$\frac{Q_2 \cdot K_2}{\sqrt{d_k}}$	$\frac{Q_3 \cdot K_2}{\sqrt{d_k}}$	$\frac{Q_4 \cdot K_2}{\sqrt{d_k}}$	$\frac{Q_5 \cdot K_2}{\sqrt{d_k}}$	\dots	$\frac{Q_n \cdot K_2}{\sqrt{d_k}}$
K_3	$\frac{Q_1 \cdot K_3}{\sqrt{d_k}}$	$\frac{Q_2 \cdot K_3}{\sqrt{d_k}}$	$\frac{Q_3 \cdot K_3}{\sqrt{d_k}}$	$\frac{Q_4 \cdot K_3}{\sqrt{d_k}}$	$\frac{Q_5 \cdot K_3}{\sqrt{d_k}}$	\dots	$\frac{Q_n \cdot K_3}{\sqrt{d_k}}$
K_4	$\frac{Q_1 \cdot K_4}{\sqrt{d_k}}$	$\frac{Q_2 \cdot K_4}{\sqrt{d_k}}$	$\frac{Q_3 \cdot K_4}{\sqrt{d_k}}$	$\frac{Q_4 \cdot K_4}{\sqrt{d_k}}$	$\frac{Q_5 \cdot K_4}{\sqrt{d_k}}$	\dots	$\frac{Q_n \cdot K_4}{\sqrt{d_k}}$
K_5	$\frac{Q_1 \cdot K_5}{\sqrt{d_k}}$	$\frac{Q_2 \cdot K_5}{\sqrt{d_k}}$	$\frac{Q_3 \cdot K_5}{\sqrt{d_k}}$	$\frac{Q_4 \cdot K_5}{\sqrt{d_k}}$	$\frac{Q_5 \cdot K_5}{\sqrt{d_k}}$	\dots	$\frac{Q_n \cdot K_5}{\sqrt{d_k}}$
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\dots	\vdots



Attention Pattern

a	fluffy	blue	creature	roamed	the	verdant	forest
\vec{E}_1	\vec{E}_2	\vec{E}_3	\vec{E}_4	\vec{E}_5	\vec{E}_6	\vec{E}_7	\vec{E}_8
\vec{Q}_1	\vec{Q}_2	\vec{Q}_3	\vec{Q}_4	\vec{Q}_5	\vec{Q}_6	\vec{Q}_7	\vec{Q}_8
$a \rightarrow \vec{E}_1 \xrightarrow{W_k} \vec{K}_1$	●	•	•	•	•	•	•
$\text{fluffy} \rightarrow \vec{E}_2 \xrightarrow{W_k} \vec{K}_2$	□	●	•	●	•	•	•
$\text{blue} \rightarrow \vec{E}_3 \xrightarrow{W_k} \vec{K}_3$	□	□	●	●	•	•	•
$\text{creature} \rightarrow \vec{E}_4 \xrightarrow{W_k} \vec{K}_4$	□	□	□	•	•	•	•
$\text{roamed} \rightarrow \vec{E}_5 \xrightarrow{W_k} \vec{K}_5$	□	□	□	□	●	•	•
$\text{the} \rightarrow \vec{E}_6 \xrightarrow{W_k} \vec{K}_6$	□	□	□	□	●	●	•
$\text{verdant} \rightarrow \vec{E}_7 \xrightarrow{W_k} \vec{K}_7$	□	□	□	□	□	●	●
$\text{forest} \rightarrow \vec{E}_8 \xrightarrow{W_k} \vec{K}_8$	□	□	□	□	□	□	•

Unnormalized Attention Pattern

+3.53	+0.80	+1.96	+4.48	+3.74	-1.95
$-\infty$	-0.30	-0.21	+0.82	+0.29	+2.91
$-\infty$	$-\infty$	+0.89	+0.67	+2.99	-0.41
$-\infty$	$-\infty$	$-\infty$	+1.31	+1.73	-1.48
$-\infty$	$-\infty$	$-\infty$	$-\infty$	+3.07	+2.94
$-\infty$	$-\infty$	$-\infty$	$-\infty$	$-\infty$	+0.31

softmax
→

Normalized Attention Pattern

1.00	0.75	0.69	0.92	0.46	0.00
0.00	0.25	0.08	0.02	0.01	0.46
0.00	0.00	0.24	0.02	0.22	0.02
0.00	0.00	0.00	0.04	0.06	0.01
0.00	0.00	0.00	0.00	0.24	0.48
0.00	0.00	0.00	0.00	0.00	0.03

Sparse Attention Mechanisms

Blockwise Attention

Linformer

Reformer

Ring attention

Longformer

Adaptive Attention Span

:

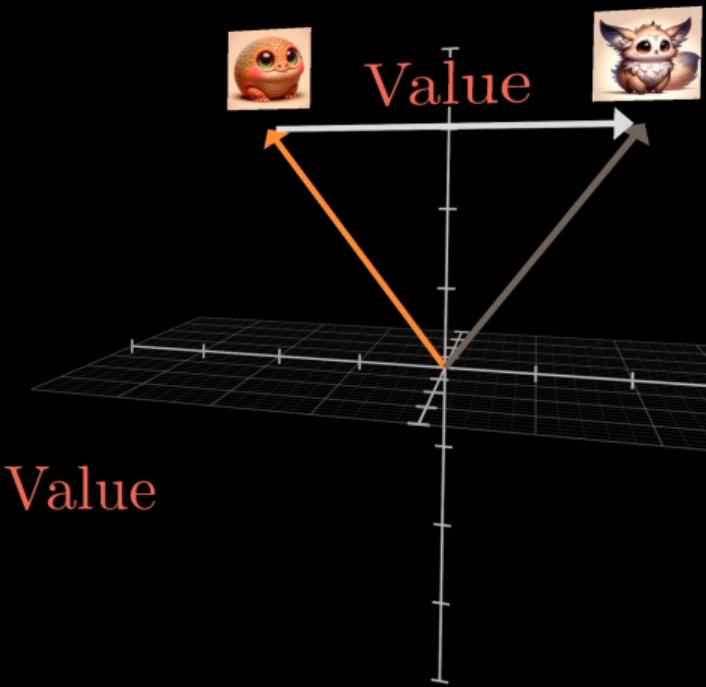
fluffy creature

$$\begin{bmatrix} +9.2 \\ -2.3 \\ +5.8 \\ +0.6 \\ +1.3 \\ +8.4 \\ \vdots \\ -8.2 \end{bmatrix}$$

$$\begin{bmatrix} -7.6 \\ +2.8 \\ -7.1 \\ +8.8 \\ +0.4 \\ -1.7 \\ \vdots \\ -4.7 \end{bmatrix}$$

W_V

$$\left[\begin{array}{ccccccccc} -3.6 & -1.7 & -8.6 & +3.8 & +1.3 & -4.6 & \cdots & -8.0 & +9.2 \\ +1.5 & +8.5 & -3.6 & +3.3 & -7.3 & +4.3 & \cdots & -6.3 & -2.3 \\ +1.7 & -9.5 & +6.5 & -9.8 & +3.5 & -4.6 & \cdots & +9.2 & +5.8 \\ -5.0 & +1.5 & +1.8 & +1.4 & -5.5 & +9.0 & \cdots & +6.9 & +0.6 \\ +3.9 & -4.0 & +6.2 & -2.0 & +7.5 & +1.6 & \cdots & +3.8 & +1.3 \\ +4.5 & +0.0 & +9.0 & +2.9 & -1.5 & +2.1 & \cdots & -3.9 & +8.4 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ +1.5 & +3.0 & +3.0 & -1.4 & +7.9 & -2.6 & \cdots & +7.8 & -8.2 \end{array} \right] = \begin{bmatrix} -52.4 \\ +89.3 \\ -80.2 \\ -17.8 \\ +7.3 \\ +223.8 \\ \vdots \\ -41.0 \end{bmatrix}$$



blue | fluffy | creature



$$\begin{bmatrix} +1.0 \\ +4.3 \\ +2.0 \\ +0.9 \\ -1.5 \\ +2.9 \\ \vdots \\ +7.8 \end{bmatrix}$$

$$\begin{bmatrix} +9.2 \\ -2.3 \\ +5.8 \\ +0.6 \\ +1.3 \\ +8.4 \\ \vdots \\ -8.2 \end{bmatrix}$$

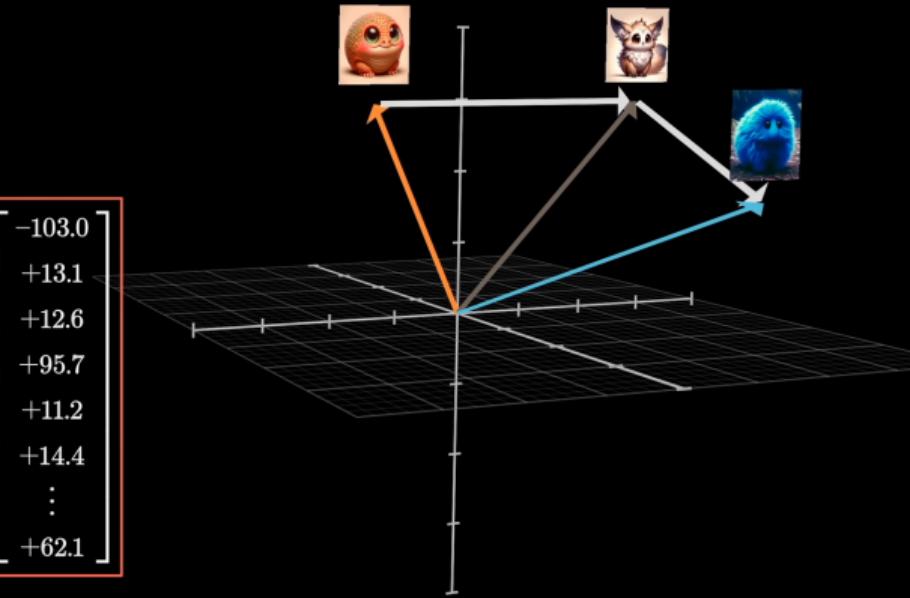
$$\begin{bmatrix} -7.6 \\ +2.8 \\ -7.1 \\ +8.8 \\ +0.4 \\ -1.7 \\ \vdots \\ -4.7 \end{bmatrix}$$

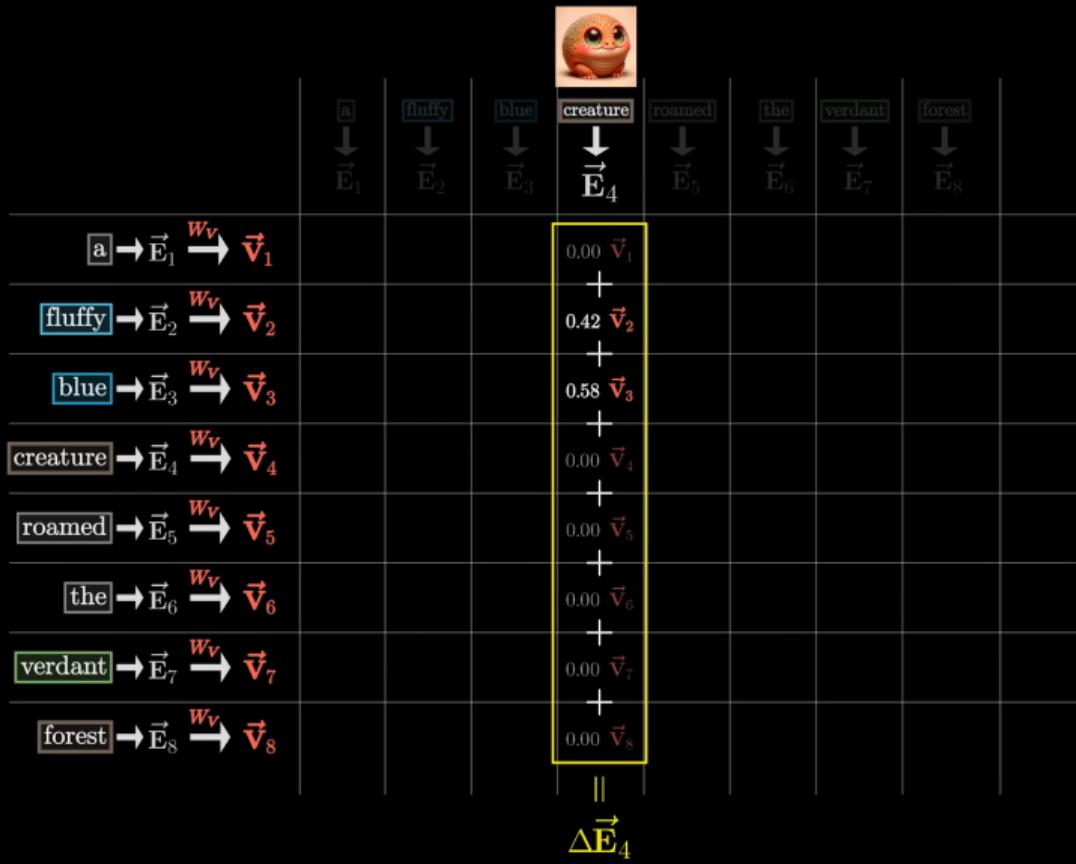
W_V

$$\begin{bmatrix} -3.6 & -1.7 & -8.6 & +3.8 & +1.3 & -4.6 & \cdots & -8.0 \\ +1.5 & +8.5 & -3.6 & +3.3 & -7.3 & +4.3 & \cdots & -6.3 \\ +1.7 & -9.5 & +6.5 & -9.8 & +3.5 & -4.6 & \cdots & +9.2 \\ -5.0 & +1.5 & +1.8 & +1.4 & -5.5 & +9.0 & \cdots & +6.9 \\ +3.9 & -4.0 & +6.2 & -2.0 & +7.5 & +1.6 & \cdots & +3.8 \\ +4.5 & +0.0 & +9.0 & +2.9 & -1.5 & +2.1 & \cdots & -3.9 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ +1.5 & +3.0 & +3.0 & -1.4 & +7.9 & -2.6 & \cdots & +7.8 \end{bmatrix} = \begin{bmatrix} +1.0 \\ +4.3 \\ +2.0 \\ +0.9 \\ -1.5 \\ +2.9 \\ \vdots \\ +7.8 \end{bmatrix}$$

$$= \begin{bmatrix} -103.0 \\ +13.1 \\ +12.6 \\ +95.7 \\ +11.2 \\ +14.4 \\ \vdots \\ +62.1 \end{bmatrix}$$

=





\boxed{a}	fluffy	blue	creature	roamed	the	verdant	forest	
\vec{E}_1	\downarrow	\vec{E}_2	\vec{E}_3	\vec{E}_4	\vec{E}_5	\vec{E}_6	\vec{E}_7	\vec{E}_8
$\vec{E}_1 \xrightarrow{w_v} \vec{v}_1$	1.00	\vec{v}_1	0.00 \vec{v}_1	0.00 \vec{v}_1	0.00 \vec{v}_1	0.00 \vec{v}_1	0.00 \vec{v}_1	0.00 \vec{v}_1
$\vec{E}_2 \xrightarrow{w_v} \vec{v}_2$	+	+	+	+	+	+	+	+
$\vec{E}_3 \xrightarrow{w_v} \vec{v}_3$	0.00 \vec{v}_3	0.00 \vec{v}_3	1.00 \vec{v}_3	0.58 \vec{v}_3	0.00 \vec{v}_3	0.00 \vec{v}_3	0.00 \vec{v}_3	0.00 \vec{v}_3
$\vec{E}_4 \xrightarrow{w_v} \vec{v}_4$	0.00 \vec{v}_4	0.00 \vec{v}_4						
$\vec{E}_5 \xrightarrow{w_v} \vec{v}_5$	0.00 \vec{v}_5	0.00 \vec{v}_5	0.00 \vec{v}_5	0.00 \vec{v}_5	0.01 \vec{v}_5	0.00 \vec{v}_5	0.00 \vec{v}_5	0.00 \vec{v}_5
$\vec{E}_6 \xrightarrow{w_v} \vec{v}_6$	0.00 \vec{v}_6	0.00 \vec{v}_6	0.00 \vec{v}_6	0.00 \vec{v}_6	0.99 \vec{v}_6	1.00 \vec{v}_6	0.00 \vec{v}_6	0.00 \vec{v}_6
$\vec{E}_7 \xrightarrow{w_v} \vec{v}_7$	0.00 \vec{v}_7	1.00 \vec{v}_7	1.00 \vec{v}_7					
$\vec{E}_8 \xrightarrow{w_v} \vec{v}_8$	0.00 \vec{v}_8	0.00 \vec{v}_8						
$\Delta \vec{E}_1$	$\Delta \vec{E}_2$	$\Delta \vec{E}_3$	$\Delta \vec{E}_4$	$\Delta \vec{E}_5$	$\Delta \vec{E}_6$	$\Delta \vec{E}_7$	$\Delta \vec{E}_8$	

$$\begin{array}{cccccccc}
\vec{E}_1 & \vec{E}_2 & \vec{E}_3 & \vec{E}_4 & \vec{E}_5 & \vec{E}_6 & \vec{E}_7 & \vec{E}_8 \\
+ & + & + & + & + & + & + & + \\
\Delta \vec{E}_1 & \Delta \vec{E}_2 & \Delta \vec{E}_3 & \Delta \vec{E}_4 & \Delta \vec{E}_5 & \Delta \vec{E}_6 & \Delta \vec{E}_7 & \Delta \vec{E}_8 \\
|| & || & || & || & || & || & || & || \\
\vec{E}'_1 & \vec{E}'_2 & \vec{E}'_3 & \vec{E}'_4 & \vec{E}'_5 & \vec{E}'_6 & \vec{E}'_7 & \vec{E}'_8
\end{array}$$

One head of attention

	a	fluffy	blue	creature	roamed	the	verdant	forest
	\vec{E}_1	\vec{E}_2	\vec{E}_3	\vec{E}_4	\vec{E}_5	\vec{E}_6	\vec{E}_7	\vec{E}_8
$\vec{E}_1 \xrightarrow{W_V} \vec{V}_1$	1.00 \vec{V}_1	0.00 \vec{V}_1						
$\vec{E}_2 \xrightarrow{W_V} \vec{V}_2$	0.00 \vec{V}_2	1.00 \vec{V}_2	0.00 \vec{V}_2	0.42 \vec{V}_2	0.00 \vec{V}_2	0.00 \vec{V}_2	0.00 \vec{V}_2	0.00 \vec{V}_2
$\vec{E}_3 \xrightarrow{W_V} \vec{V}_3$	0.00 \vec{V}_3	0.00 \vec{V}_3	1.00 \vec{V}_3	0.58 \vec{V}_3	0.00 \vec{V}_3	0.00 \vec{V}_3	0.00 \vec{V}_3	0.00 \vec{V}_3
$\vec{E}_4 \xrightarrow{W_V} \vec{V}_4$	0.00 \vec{V}_4	0.00 \vec{V}_4	0.00 \vec{V}_4	0.00 \vec{V}_4	1.00 \vec{V}_4	0.00 \vec{V}_4	0.00 \vec{V}_4	0.00 \vec{V}_4
$\vec{E}_5 \xrightarrow{W_V} \vec{V}_5$	0.00 \vec{V}_5	0.00 \vec{V}_5	0.00 \vec{V}_5	0.00 \vec{V}_5	0.01 \vec{V}_5	0.00 \vec{V}_5	0.00 \vec{V}_5	0.00 \vec{V}_5
$\vec{E}_6 \xrightarrow{W_V} \vec{V}_6$	0.00 \vec{V}_6	0.00 \vec{V}_6	0.00 \vec{V}_6	0.00 \vec{V}_6	0.99 \vec{V}_6	1.00 \vec{V}_6	0.00 \vec{V}_6	0.00 \vec{V}_6
$\vec{E}_7 \xrightarrow{W_V} \vec{V}_7$	0.00 \vec{V}_7	1.00 \vec{V}_7	1.00 \vec{V}_7					
$\vec{E}_8 \xrightarrow{W_V} \vec{V}_8$	0.00 \vec{V}_8							
	$\Delta \vec{E}_1$	$\Delta \vec{E}_2$	$\Delta \vec{E}_3$	$\Delta \vec{E}_4$	$\Delta \vec{E}_5$	$\Delta \vec{E}_6$	$\Delta \vec{E}_7$	$\Delta \vec{E}_8$
	\vec{E}'_1	\vec{E}'_2	\vec{E}'_3	\vec{E}'_4	\vec{E}'_5	\vec{E}'_6	\vec{E}'_7	\vec{E}'_8
	\vec{E}_1'	\vec{E}_2'	\vec{E}_3'	\vec{E}_4'	\vec{E}_5'	\vec{E}_6'	\vec{E}_7'	\vec{E}_8'
	+	+	+	+	+	+	+	+
	$\Delta \vec{E}_1'$	$\Delta \vec{E}_2'$	$\Delta \vec{E}_3'$	$\Delta \vec{E}_4'$	$\Delta \vec{E}_5'$	$\Delta \vec{E}_6'$	$\Delta \vec{E}_7'$	$\Delta \vec{E}_8'$

Query

Key

$$128 \left\{ \begin{bmatrix} -3.7 & +3.9 & -2.4 & -6.3 & -9.4 & -8.6 & +3.6 & -0.9 & \cdots & +0.7 \\ +7.9 & +9.7 & -5.6 & +3.2 & -4.7 & -9.5 & +5.1 & -3.6 & \cdots & -2.3 \\ +1.7 & +6.6 & +2.6 & +7.4 & -4.5 & +5.9 & -6.2 & +9.0 & \cdots & +3.7 \\ \vdots & \ddots & \vdots \\ -5.6 & +8.9 & +4.6 & -4.9 & -5.7 & +0.4 & -9.4 & -5.8 & \cdots & -1.5 \end{bmatrix} \right\}$$

12,288

$$128 \left\{ \begin{bmatrix} -2.5 & -0.7 & -4.4 & +1.7 & +7.2 & -7.6 & +0.3 & -7.3 & \cdots & +4.3 \\ -2.1 & +1.3 & -6.3 & -7.0 & -0.2 & -2.9 & +8.7 & +5.3 & \cdots & +4.9 \\ +8.0 & -8.2 & +1.0 & +1.7 & +9.1 & -4.1 & -5.1 & -7.9 & \cdots & -9.6 \\ \vdots & \ddots & \vdots \\ +8.5 & +3.4 & +5.6 & -4.3 & +1.7 & -8.6 & -0.3 & +9.5 & \cdots & +7.5 \end{bmatrix} \right\}$$

12,288

Value



Value

$$12,288 \times 12,288 = 150,994,944$$

Query
1,572,864

$$\begin{bmatrix} -3.7 & +3.9 & -2.4 & -6.3 & -9.4 & -8.6 & +3.6 & -0.9 & \cdots & +0.7 \\ +7.9 & +9.7 & -5.6 & +3.2 & -4.7 & -9.5 & +5.1 & -3.6 & \cdots & -2.3 \\ +1.7 & +6.6 & +2.6 & +7.4 & -4.5 & +5.9 & -6.2 & +9.0 & \cdots & +3.7 \\ \vdots & \ddots & \vdots \\ -5.6 & +8.9 & +4.6 & -4.9 & -5.7 & +0.4 & -9.4 & -5.8 & \cdots & -1.5 \end{bmatrix}$$

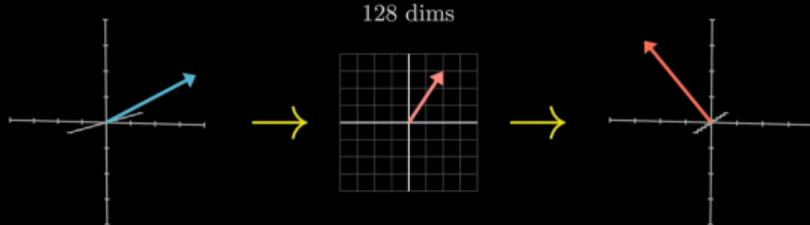
Key
1,572,864

$$\begin{bmatrix} -2.5 & -0.7 & -4.4 & +1.7 & +7.2 & -7.6 & +0.3 & -7.3 & \cdots & +4.3 \\ -2.1 & +1.3 & -6.3 & -7.0 & -0.2 & -2.9 & +8.7 & +5.3 & \cdots & +4.9 \\ +8.0 & -8.2 & +1.0 & +1.7 & +9.1 & -4.1 & -5.1 & -7.9 & \cdots & -9.6 \\ \vdots & \ddots & \vdots \\ +8.5 & +3.4 & +5.6 & -4.3 & +1.7 & -8.6 & -0.3 & +9.5 & \cdots & +7.5 \end{bmatrix}$$

$$12,288 \underbrace{\begin{bmatrix} -3.2 & +9.1 & -5.3 & +8.9 & +8.7 & +5.9 & +2.6 & +7.4 & \cdots & -4.1 \\ +6.9 & +2.3 & -9.6 & -3.0 & -7.0 & +9.5 & -0.4 & -0.1 & \cdots & +2.8 \\ -2.6 & -7.2 & +6.4 & -6.1 & +0.2 & -5.5 & -8.0 & +7.2 & \cdots & +9.4 \\ +9.1 & +8.0 & +5.4 & -3.3 & -8.3 & -1.8 & -5.3 & -7.3 & \cdots & -8.8 \\ +4.5 & -9.7 & +5.4 & -7.0 & -8.3 & -8.1 & +3.4 & -5.0 & \cdots & -1.6 \\ +1.1 & +7.1 & +4.5 & -4.5 & -7.3 & -8.8 & -3.9 & -4.7 & \cdots & -0.9 \\ +3.6 & +3.9 & -4.3 & -2.4 & -6.3 & +5.7 & -8.8 & +3.9 & \cdots & +5.5 \\ +5.5 & -4.8 & -2.5 & +1.7 & -4.5 & -2.6 & -6.0 & -0.8 & \cdots & -9.0 \\ \vdots & \ddots & \vdots \\ +5.9 & -8.4 & +0.4 & -3.8 & +1.5 & +9.1 & +2.9 & -9.2 & \cdots & -1.4 \end{bmatrix}}_{12,288} = \begin{bmatrix} +0.2 \\ +73.1 \\ -28.2 \\ +119.4 \\ +215.7 \\ +91.8 \\ -29.1 \\ -5.6 \\ \vdots \\ -5.1 \end{bmatrix}$$

12,288 dims

12,288 dims



“Low rank” transformation

d_input	d_output
12,288	12,288

$$\begin{matrix}
 & \overbrace{\hspace{10em}}^{128} \\
 12,288 & \left\{ \begin{matrix}
 \begin{bmatrix}
 +5.0 & -3.3 & \cdots & +7.2 \\
 -8.9 & -4.9 & \cdots & -7.8 \\
 -3.0 & +4.8 & \cdots & +2.4 \\
 +4.2 & -5.8 & \cdots & +3.5 \\
 +7.5 & +0.9 & \cdots & -9.3 \\
 +4.2 & -9.7 & \cdots & +0.6 \\
 +8.4 & -8.1 & \cdots & -9.4 \\
 -3.1 & +2.4 & \cdots & -5.7 \\
 \vdots & \vdots & \ddots & \vdots \\
 +8.9 & -9.8 & \cdots & +2.0
 \end{bmatrix} &
 \begin{bmatrix}
 -7.8 & +8.9 & -5.3 & +3.8 & -8.7 & +4.6 & +7.6 & -4.5 & \cdots & -2.5 \\
 +4.9 & -5.2 & -6.5 & -1.0 & -3.9 & +6.7 & -5.2 & +0.0 & \cdots & +2.7 \\
 +7.3 & +8.7 & +5.0 & +4.0 & +9.3 & +9.8 & -1.0 & -8.5 & \cdots & -6.9 \\
 \vdots & \ddots & \vdots \\
 -1.8 & +1.0 & -4.5 & -0.9 & -1.9 & -5.0 & +0.1 & -3.8 & \cdots & +0.5
 \end{bmatrix} \\
 & \overbrace{\hspace{10em}}^{12,288}
 \end{matrix}
 \right. \\
 & \text{Value}_\downarrow \\
 & \text{Value}_\uparrow
 \end{matrix}$$

\downarrow

$=$

\downarrow

$\begin{bmatrix} +0.2 \\ +0.7 \\ +3.6 \\ -4.4 \\ -7.3 \\ -2.1 \\ +9.0 \\ -29.1 \\ -6.2 \\ \vdots \\ +0.9 \end{bmatrix}$

$=$

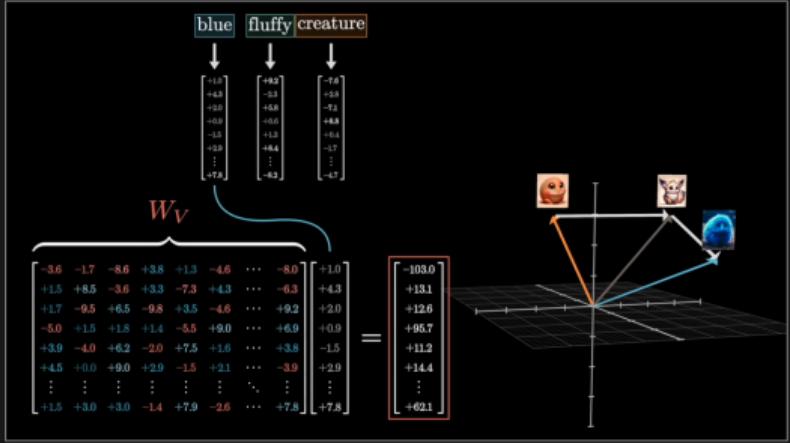
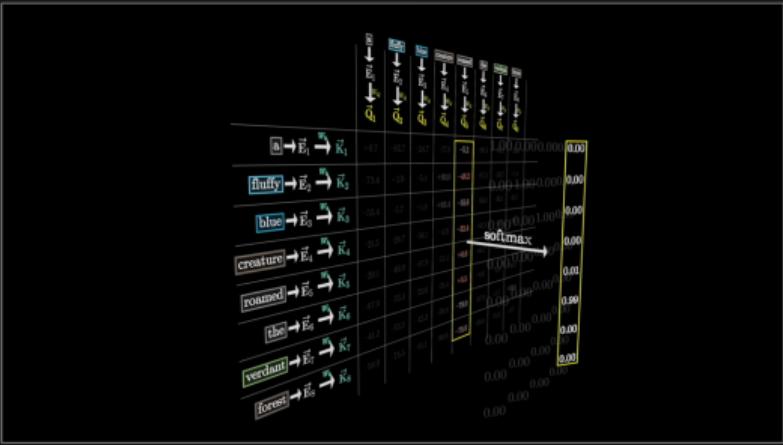
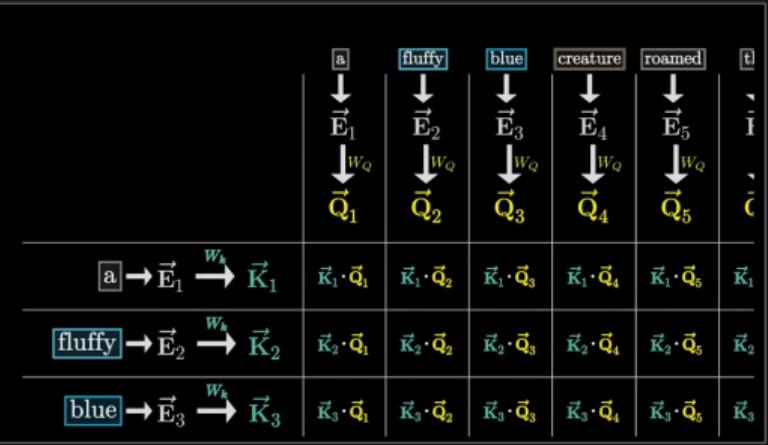
$\begin{bmatrix} -198.6 \\ +73.1 \\ -28.2 \\ +119.4 \\ +215.7 \\ +91.8 \\ -29.1 \\ -5.6 \\ \vdots \\ -5.1 \end{bmatrix}$

Total weights: 175,181,291,520

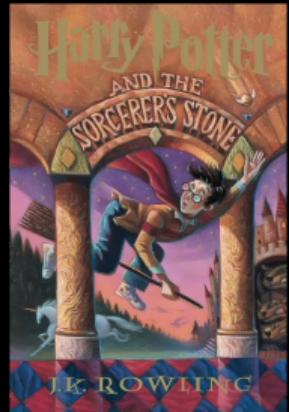
Organized into 27,938 matrices



Embedding	12,288 d_embed * n_vocab = 617,558,016
Key	128 d_query * d_embed = 1,572,864 per head
Query	128 d_query * d_embed = 1,572,864 per head
Value _↓	128 d_value * d_embed = 1,572,864 per head
Value _↑	12,288 d_embed * d_value = 1,572,864 per head
Up-projection	6,291,456
Down-projection	
Unembedding	50,257 n_vocab * d_embed = 617,558,016



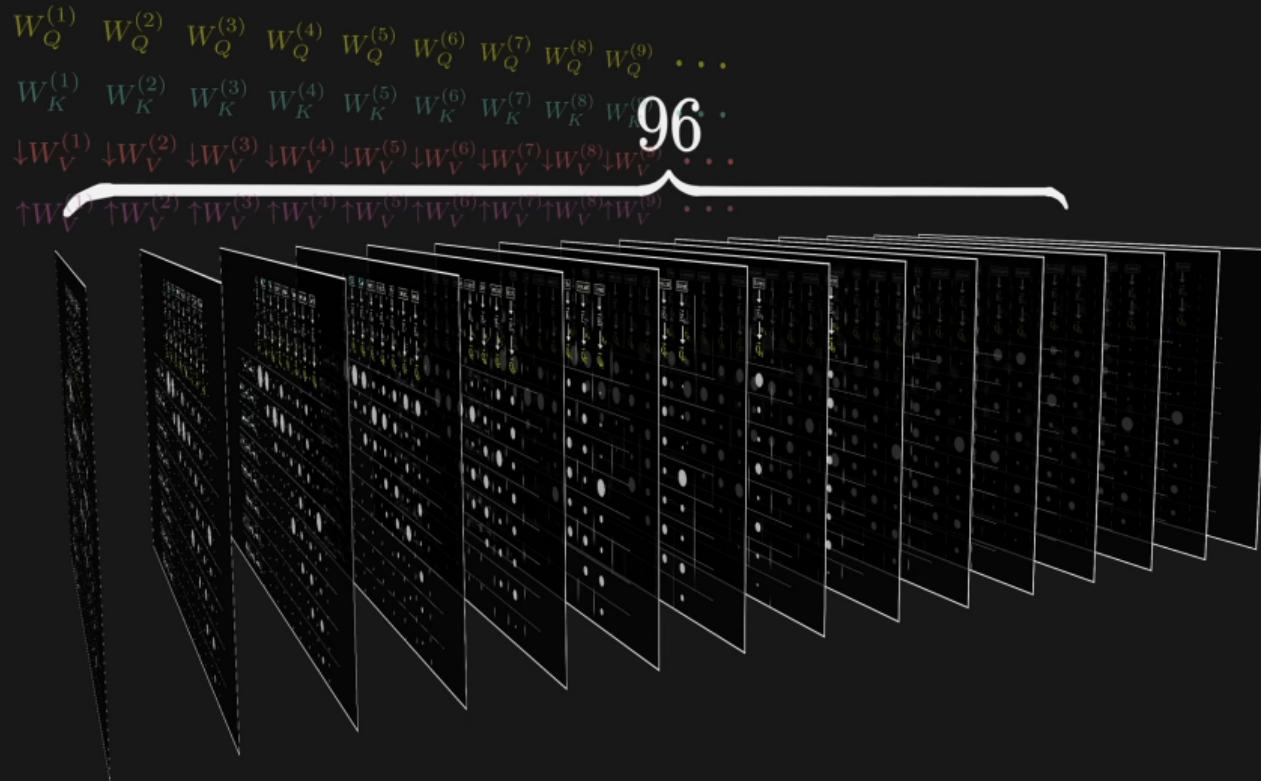
... wizard ... Hogwarts ... Hermione ... Harry

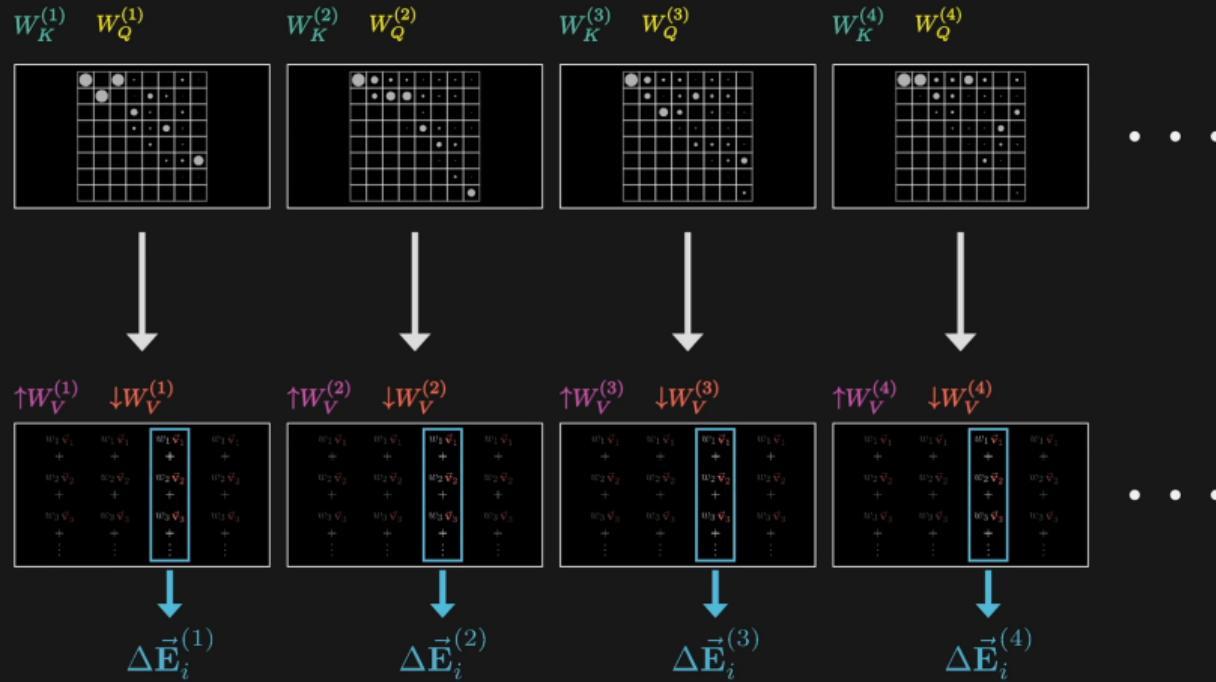


... Queen ... Sussex ... William ... Harry



Multi-headed attention





New
embedding $\boxed{\vec{E}_i + \Delta \vec{E}_i^{(1)} + \Delta \vec{E}_i^{(2)} + \Delta \vec{E}_i^{(3)} + \Delta \vec{E}_i^{(4)} + \dots}$

Total weights: 175,181,291,520

Organized into 27,938 matrices



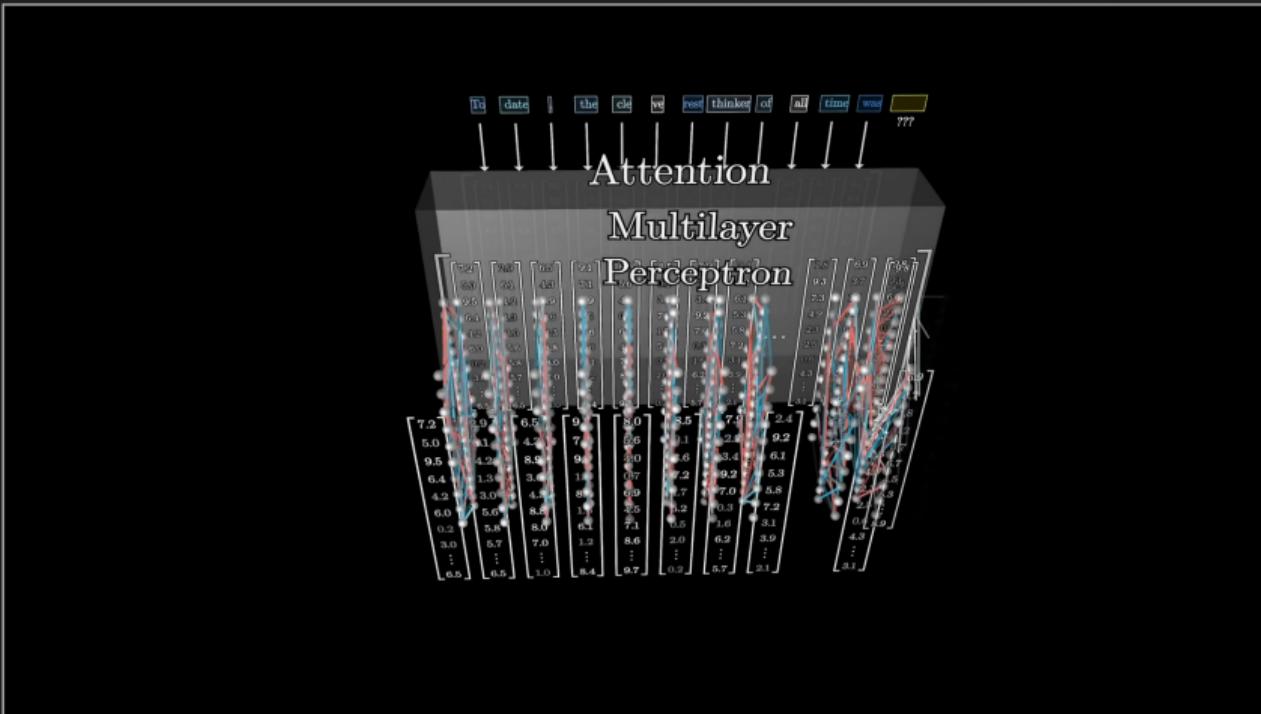
Embedding	$d_{\text{embed}} \times n_{\text{vocab}}$	= 617,558,016	
Key	$d_{\text{query}} \times d_{\text{embed}} \times n_{\text{heads}}$	= 150,994,944	per layer
Query	$d_{\text{query}} \times d_{\text{embed}} \times n_{\text{heads}}$	= 150,994,944	per layer
Value _↓	$d_{\text{value}} \times d_{\text{embed}} \times n_{\text{heads}}$	= 150,994,944	per layer
Value _↑	$d_{\text{embed}} \times d_{\text{value}} \times n_{\text{heads}}$	= 150,994,944	per layer
Up-projection		603,979,776	
Down-projection			
Unembedding	$n_{\text{vocab}} \times d_{\text{embed}}$	= 617,558,016	

Total weights: 175,181,291,520
Organized into 27,938 matrices

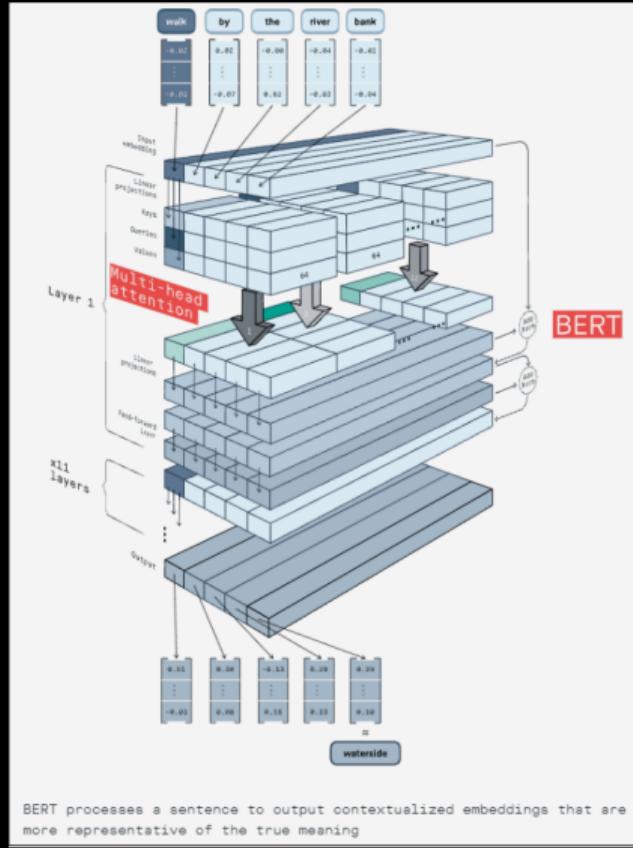


Embedding	12,288	50,257	$d_{\text{embed}} * n_{\text{vocab}} = 617,558,016$	
Key	128	12,288	96	96
Query	128	12,288	96	96
Value	128	12,288	96	96
Output	12,288	128	96	96
Up-projection				57,982,058,496
Down-projection				
Unembedding	50,257	12,288	$n_{\text{vocab}} * d_{\text{embed}} = 617,558,016$	

About 1/3 of What Attention is ~~All~~ You Need



En 2020 : on avait que BERT! (version française)



Un exemple de la corvée du Dr Brunet-Gouet

jf+, js+ J'ai 2 frères et deux sœurs.

& Je m'entend plutôt bien avec.

mp- Ma maman est séparée de mon père.

jm+ Je m'entend très bien avec ma mère.

si J'ai deux grandes demi-sœur du côté de mon père.

& Elles ont déjà des enfants.

mi,pi Ma maman à fait cinq enfants, mon père en n'a fait trois.

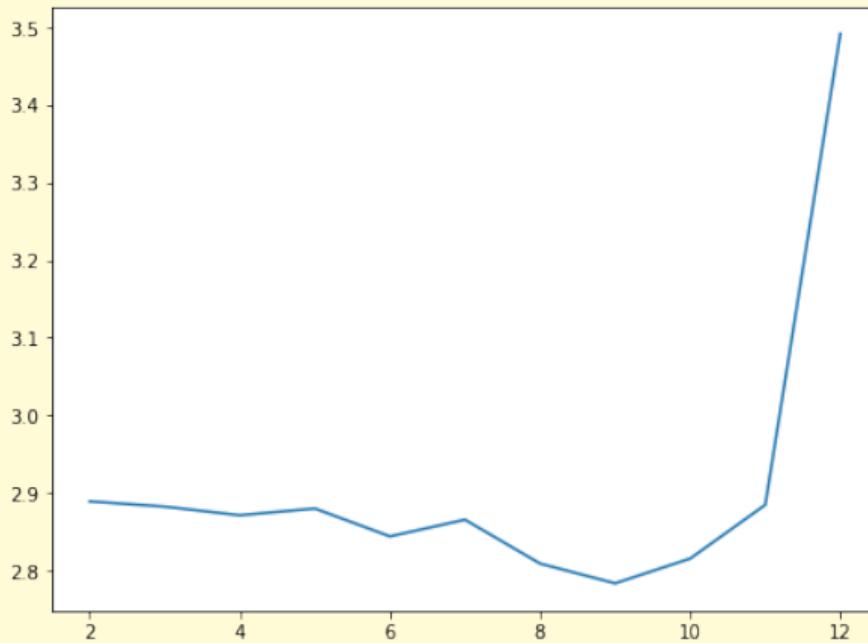
jm+ Avec ma mère ont aime bien faire les magasins.

Bilan : 1648 phrases et 11 labels

- La valence relationnelle : **+, - 0**. Les relations positives font référence à une bonne entente, à l'expression d'un affect positif et à la coopération. Les relations négatives correspondent à des conflits, des désaccords, l'absence d'une relation normale.
- En l'absence d'information sur la valence, le texte est considéré comme informatif (**i**) sur les habitudes ou les conditions de vie des personnes.
- Les sujets décrits dans un segment ont été étiquetés comme suit : le répondant **je**, la **mère**, le **père**, la **soeur**, le **frère**, la **famille** et une **tierce personne**.

Partition apprentissage-test de tailles 1318 et 330.

10 jours de calcul sur NVIDIA A100 Tensor Core 40GB GPU



Résultats numériques

Precision = TP/TP+FP, Recall = TP/TP+FN et F1 Score = $2 * (\text{Recall} * \text{Precision}) / (\text{Recall} + \text{Precision})$

Label	+	-	0	i	j	f	s	p	m	a	t
Support	72	63	47	146	263	36	57	124	123	72	23
Precision											
Fine-tuning	0.67	0.85	0.40	0.78	0.86	0.91	0.97	0.97	0.94	0.91	0.78
Elasticnet lr	0.65(0.69)	0.71(0.75)	0.28(0.33)	0.74(0.61)	0.85(0.81)	0.74(0.79)	0.89(0.98)	0.90(0.93)	0.83(0.87)	0.77(0.81)	0.67(0.75)
Gradient Boosting	0.73(0.65)	0.88(0.44)	0.25(0.45)	0.75(0.64)	0.84(0.80)	0.75(0.78)	0.88(0.92)	0.84(0.93)	0.79(0.85)	0.83(0.73)	0(0.83)
Random Forest	0.82(0.70)	0.71(0.73)	0(0.33)	0.71(0.64)	0.82(0.80)	0(0.72)	0.89(0.92)	0.86(0.93)	0.76(0.89)	0.73(0.77)	0(0.67)
SVC	0.68(0.70)	0.79(1)	0(0)	0.76(0.66)	0.85(0.81)	0.68(0.79)	0.91(0.94)	0.90(0.92)	0.83(0.89)	0.82(0.79)	0.50(0.79)
Recall											
Fine-tuning	0.82	0.37	0.38	0.81	0.97	0.89	0.98	0.93	0.93	0.86	0.61
Elasticnet lr	0.56(0.49)	0.40(0.10)	0.11(0.02)	0.73(0.75)	0.94(1)	0.39(0.64)	0.68(0.81)	0.76(0.85)	0.82(0.85)	0.67(0.64)	0.26(0.52)
Gradient Boosting	0.49(0.49)	0.22(0.11)	0.02(0.11)	0.75(0.66)	0.97(0.95)	0.17(0.81)	0.49(0.86)	0.70(0.85)	0.69(0.90)	0.47(0.64)	0.00(0.43)
Random Forest	0.38(0.43)	0.08(0.17)	0(0.04)	0.75(0.59)	0.96(0.92)	0.00(0.86)	0.30(0.95)	0.52(0.92)	0.60(0.93)	0.33(0.71)	0.00(0.52)
SVC	0.57(0.46)	0.30(0.05)	0(0)	0.77(0.73)	0.94(1)	0.42(0.64)	0.72(0.86)	0.77(0.78)	0.81(0.89)	0.64(0.58)	0.09(0.48)
f1-score											
Fine-tuning	0.74	0.51	0.39	0.79	0.91	0.90	0.97	0.95	0.93	0.89	0.68
Elasticnet lr	0.60(0.57)	0.51(0.17)	0.15(0.04)	0.74(0.67)	0.89(0.89)	0.51(0.71)	0.77(0.88)	0.82(0.89)	0.83(0.86)	0.72(0.71)	0.38(0.62)
Gradient Boosting	0.58(0.56)	0.35(0.18)	0.04(0.17)	0.75(0.65)	0.90(0.87)	0.27(0.79)	0.63(0.89)	0.77(0.89)	0.74(0.88)	0.60(0.68)	0(0.57)
Random Forest	0.51(0.53)	0.14(0.28)	0(0.08)	0.73(0.61)	0.89(0.86)	0.00(0.78)	0.45(0.93)	0.65(0.92)	0.67(0.91)	0.46(0.74)	0(0.59)
SVC	0.62(0.55)	0.44(0.09)	0(0)	0.77(0.69)	0.90(0.89)	0.52(0.71)	0.80(0.90)	0.83(0.84)	0.82(0.89)	0.72(0.67)	0.15(0.59)

Un exemple de tokenizer : Byte-Pair encoding

```
corpus = [
    "This example is taken from the multitude of tokenizer examples
     explained on the Hugging Face courses.",
    "This is one example of tokenization.",
    "There is several tokenizer algorithms.",
    "Hopefully, you will be able to understand how they are trained and generate tokens.",
]

{'This': 2, 'example': 2, 'is': 3, 'taken': 1, 'from': 1, 'the': 2,
'multitude': 1, 'of': 2, 'tokenizer': 2, 'examples': 1, 'explained': 1,
'on': 1, 'Hugging': 1, 'Face': 1, 'courses': 1, '.': 4, 'one': 1,
'tokenization': 1, 'There': 1, 'several': 1, 'algorithms': 1,
'Hopefully': 1, ',': 1, 'you': 1, 'will': 1, 'be': 1, 'able': 1,
'to': 1, 'understand': 1, 'how': 1, 'they': 1, 'are': 1,
'trained': 1, 'and': 1, 'generate': 1, 'tokens': 1}
```

Un exemple de tokenizer : Byte-Pair encoding

En partant d'un alphabet

```
[',', '.', 'F', 'H', 'T', 'a', 'b', 'c', 'd', 'e', 'f', 'g', 'h', 'i', 'k',  
'l', 'm', 'n', 'o', 'p', 'r', 's', 't', 'u', 'v', 'w', 'x', 'y', 'z', 'G']
```

On découpe les mots comme suit

```
{'This': ['T', 'h', 'i', 's'],  
'Gexample': ['G', 'e', 'x', 'a', 'm', 'p', 'l', 'e'],  
'Gis': ['G', 'i', 's'],  
'Gtaken': ['G', 't', 'a', 'k', 'e', 'n'],  
'Gfrom': ['G', 'f', 'r', 'o', 'm'],  
'Gthe': ['G', 't', 'h', 'e'],  
'Gmultitude': ['G', 'm', 'u', 'l', 't', 'i', 't', 'u', 'd', 'e'],  
'Gof': ['G', 'o', 'f'],  
'Gtokenizer': ['G', 't', 'o', 'k', 'e', 'n', 'i', 'z', 'e', 'r'],  
'Gexamples': ['G', 'e', 'x', 'a', 'm', 'p', 'l', 'e', 's'],  
'Gexplained': ['G', 'e', 'x', 'p', 'l', 'a', 'i', 'n', 'e', 'd'], ...}
```

Un exemple de tokenizer : Byte-Pair encoding

On va compter les paires de lettres de l'alphabet

```
('T', 'h'): 3
('h', 'i'): 2
('i', 's'): 5
('G', 'e'): 4
('e', 'x'): 4
('x', 'a'): 3
('a', 'm'): 3
('m', 'p'): 3
('p', 'l'): 4
('l', 'e'): 4
('G', 'i'): 3
('G', 't'): 10
.
.
```

Mise à jour de l'alphabet

```
[', '.', 'F', 'H', 'T', 'a', 'b', 'c', 'd', 'e', 'f', 'g', 'h', 'i',
'k', 'l', 'm', 'n', 'o', 'p', 'r', 's', 't', 'u', 'v', 'w', 'x', 'y',
'z', 'G', 'Gt']
```

Un exemple de tokenizer : Byte-Pair encoding

On refait la découpe des mots en fonction des éléments de l'alphabet,

```
{'Gtrained': ['G', 't', 'r', 'a', 'i', 'n', 'e', 'd'],
 'Gtokens': ['G', 't', 'o', 'k', 'e', 'n', 's']}
```

deviennent

```
{'Gtrained': ['Gt', 'r', 'a', 'i', 'n', 'e', 'd'],
 'Gtokens': ['Gt', 'o', 'k', 'e', 'n', 's']}
```

Byte-Pair encoding : procédé itératif

Critère d'arrêt : la taille de vocabulaire.

```
[ '<|endoftext|>', ',', '.', 'F', 'H', 'T', 'a', 'b',
 'c', 'd', 'e', 'f', 'g', 'h', 'i', 'k', 'l', 'm',
 'n', 'o', 'p', 'r', 's', 't', 'u', 'v', 'w', 'x',
 'y', 'z', 'G', 'Gt', 'en', 'er', 'is', 'ken', 'Gto',
 'Ge', 'Gex', 'pl', 'Go', 'Gtoken', 'Ga', 'Th', 'Gexa',
 'Gexam', 'Gexampl', 'Gexample', 'Gis', 'Gth']
```

On peut tokeniser un texte
en

```
[ 'Th', 'is', 'Gis', 'G', 'n', 'o', 't', 'Ga', 'Gtoken', '.']
```

- Un processus de conversion d'un texte en tokens individuels (mots, sous-mots ou caractères).
- Décomposer le texte en unités exploitables pour le traitement.
- Tokenizers : WordPiece, Byte Pair Encoding, SentencePiece.
- Peut varier considérablement d'une langue à l'autre en raison des différentes structures de jetons.

Que peut-on faire avec tokens ?

Un embedding (plongement) : exemple

First Citizen:

Before we proceed any further, hear me speak.

All:

Speak, speak.

First Citizen:

You are all resolved rather to die than to famish?

All:

Resolved. resolved.

First Citizen:

First, you know Caius Marcius is chief enemy to the people.

All:

We know't, we know't.

First Citizen:

Let us kill him, and we'll have corn at our own price.

Is't a verdict?

Embedding : exemple

Un tokenizer basique

On obtient l'alphabet :

!\$&',-.3:;?ABCDEFGHIJKLMNPQRSTUVWXYZabcdefghijklmnopqrstuvwxyz

65

On s'en sert comme suit : Nous donne :

[46, 47, 47, 1, 58, 46, 43, 56, 43]

hi there

Un bloc de tokens pour prédire le suivant

Avec cette séquence de tokens

```
tensor([18, 47, 56, 57, 58, 1, 15, 47, 58])
```

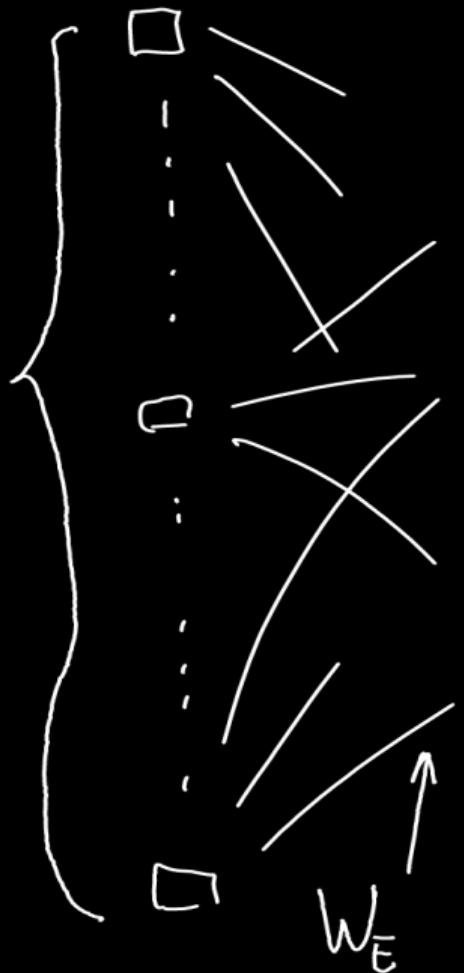
Nous allons construire ce jeu de données d'apprentissage

```
when input is tensor([18]) the target: 47
when input is tensor([18, 47]) the target: 56
when input is tensor([18, 47, 56]) the target: 57
when input is tensor([18, 47, 56, 57]) the target: 58
when input is tensor([18, 47, 56, 57, 58]) the target: 1
when input is tensor([18, 47, 56, 57, 58, 1]) the target: 15
when input is tensor([18, 47, 56, 57, 58, 1, 15]) the target: 47
when input is tensor([18, 47, 56, 57, 58, 1, 15, 47]) the target: 58
```

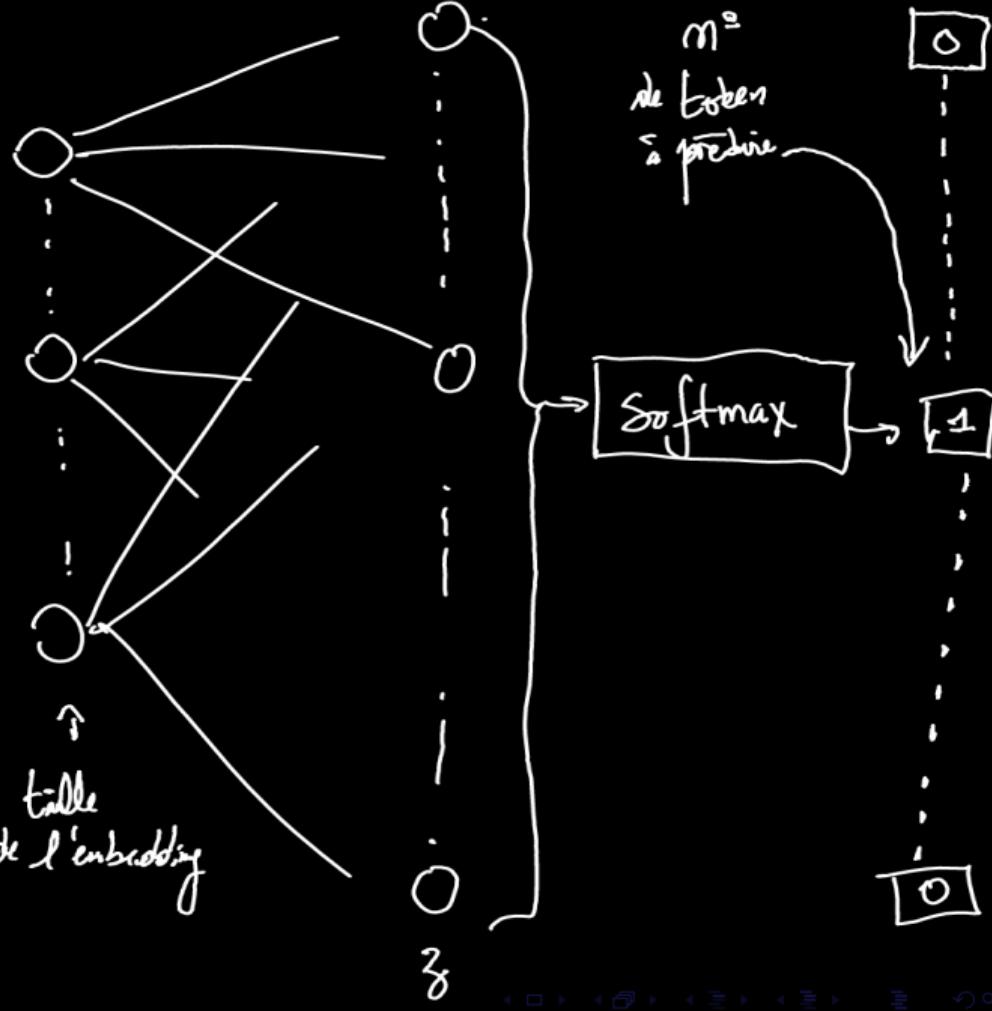
Il nous reste qu'à encoder les vecteurs et utiliser un réseau de neurones de classification.

Séquence de tokens : $(0, \dots, 2, \dots, 1, \dots, 0)$

taille de l'alphabet : langage



taille de l'embedding



taille de l'ensemble à prédire

- On récupère la matrice des poids du réseau de neurones
- Pour l'instant l'ordre des mots n'est pas pris en compte
- Pour prendre en compte l'ordre des mots : on utilise une astuce dite de *Positional Encoding*

Attention : estimation non-paramétrique

Estimateur à noyaux de Nadaraya et Watson en 1964. Pondérer les y_i en fonction de l'emplacement des x_i :

$$f(x) = \sum_{i=1}^n \frac{K(x - x_i)}{\sum_{j=1}^n K(x - x_j)} y_i,$$

où K est noyau. On peut généraliser cet estimateur à

$$f(x) = \sum_{i=1}^n \alpha(x, x_i) y_i,$$

où x est la requête et (x_i, y_i) la paire clé-valeur. La mise en commun de l'attention ici est une moyenne pondérée des valeurs y_i . Le *poids d'attention* $\alpha(x, x_i)$ est attribué à la valeur correspondante y_i sur la base de l'interaction entre la requête x et la clé x_i , modélisée par α .

Attention : estimation non-paramétrique

Pour mieux comprendre la mise en commun de l'attention, il suffit de considérer un *noyau gaussien* défini comme suit

$$K(u) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{u^2}{2}\right).$$

On obtient

$$\begin{aligned} f(x) &= \sum_{i=1}^n \alpha(x, x_i) y_i \\ &= \sum_{i=1}^n \frac{\exp\left(-\frac{1}{2}(x - x_i)^2\right)}{\sum_{j=1}^n \exp\left(-\frac{1}{2}(x - x_j)^2\right)} y_i \\ &= \sum_{i=1}^n \text{softmax}\left(-\frac{1}{2}(x - x_i)^2\right) y_i. \end{aligned}$$