

Self-Attention

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- For any NLP task , we give the input a set of words \rightarrow numbers.

- How to convert words in number called **vectorization**

- First step in NLP

- Convert the word into number

- One-hot-encoding

	mat	cat	rat
mat	1	0	0
cat	0	1	0
rat	0	0	1

OHE $\begin{bmatrix} 1) & \underline{\text{mat}} & \underline{\text{cat}} & \underline{\text{mat}} \\ 2) & \underline{\text{cat}} & \underline{\text{rat}} & \underline{\text{rat}} \end{bmatrix}$

- Problem:

num $\rightarrow [1 \ 0 \ 0] \ [0 \ 1 \ 0] \ [1 \ 0 \ 0]$

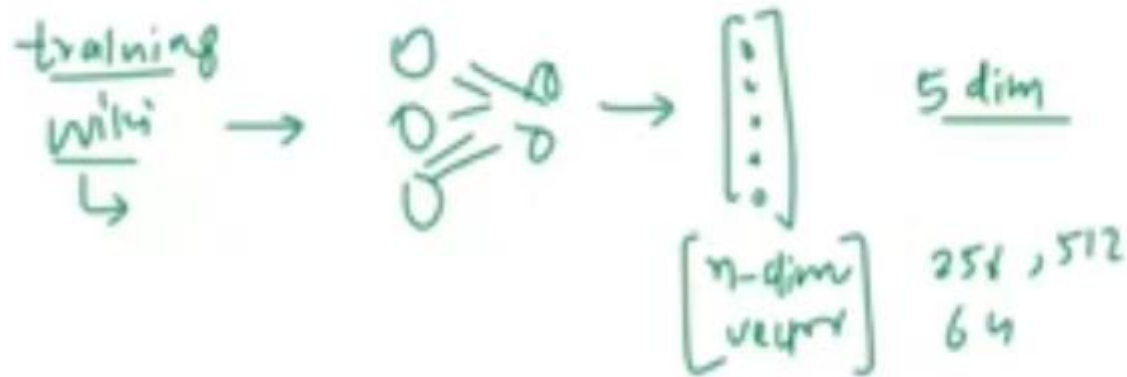
king $\rightarrow [0.6 \ 0.1 \ 1 \ 0 \ 0.9]$
queen $\rightarrow [0.3 \ 0.2 \ 0.4 \ 1 \ 0]$

- BoW(how many time a word came)

<u>BoW</u>	mat	rat	cat
<u>s1</u>	[2	0	1]
<u>s2</u>	→ [0	2	1]

- Tf-Idf

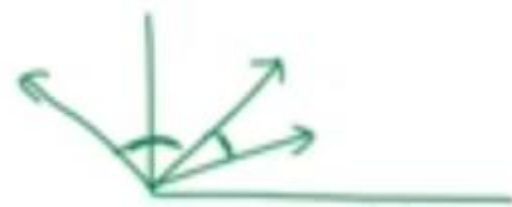
- Word Embedding (Most Powerful having Semantics)



- Pick the data send them in NN and convert it into n-dimension vector, let say 5 dimension
- Each dimension represent a factor. Each dimension has meaning.

- king $\rightarrow [0.6 \ 0.1 \ 1 \ 0 \ 0.9]$
queen $\rightarrow [0.3 \ 0.2 \ 0.4 \ 1 \ 0]$

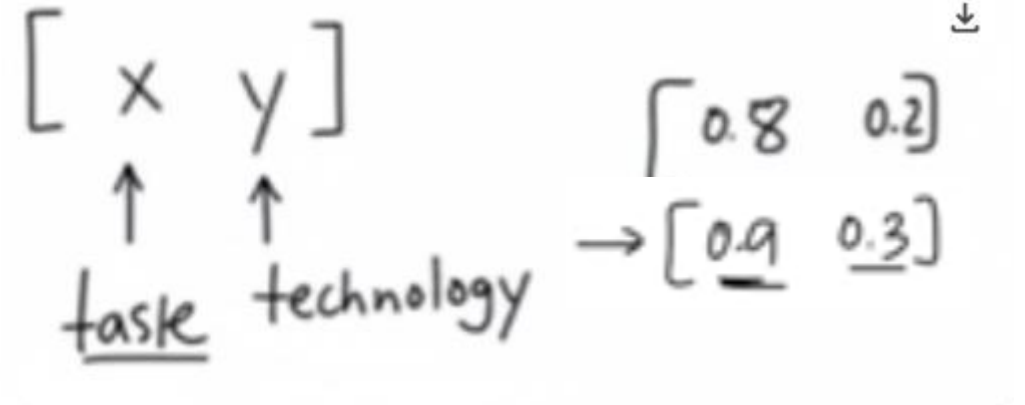
cricket $[0.2 \ 0.9 \ 0.9 \ 0.9 \ 0.4]$



Problem with w2vec Embedding:

- Capture average Meaning:

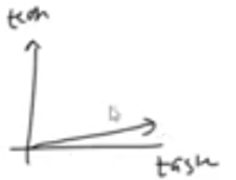
1. **An apple a day keeps the doctor away**
2. Apple is healthy
3. Apple is better than orange
4. Apple moves great phones



- Capture average meaning.
- Word Embedding \rightarrow Static Make once and used again & again
- Make once and used again & again
- It should be dynamic

Fruit \rightarrow [**Apple Launched a new Phone while I was eating an orange**]

static \rightarrow [0.9 0.3]



- Self-attention produce dynamic embedding from static embedding

- Static embedding is:

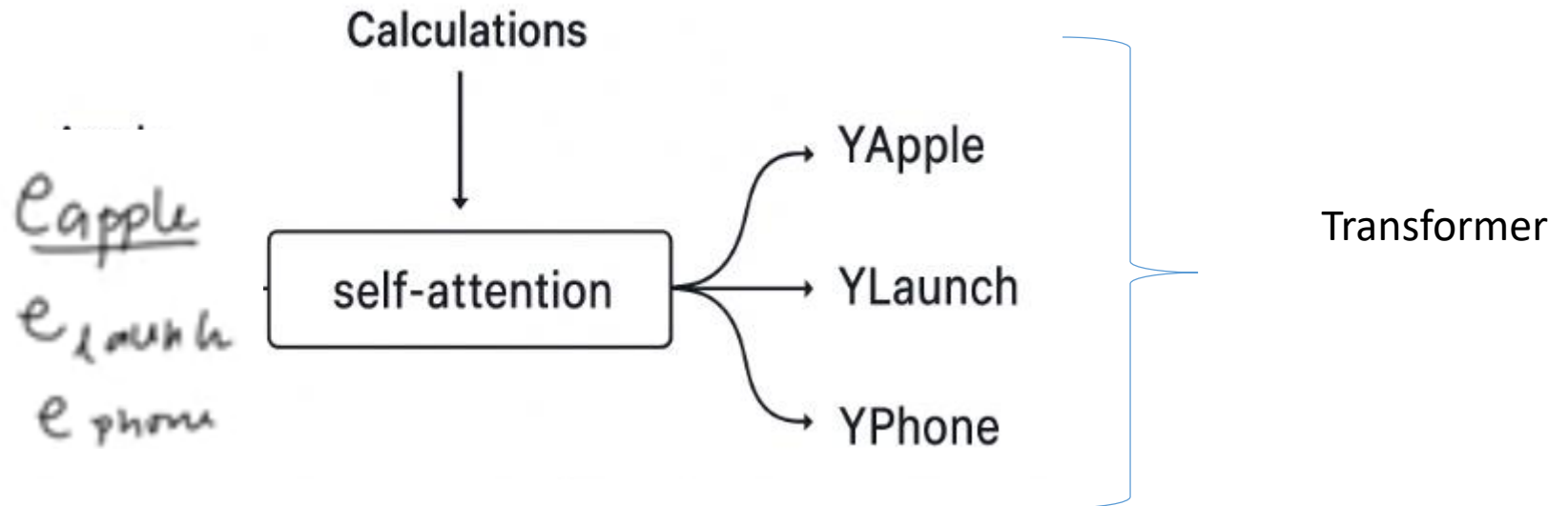
$$\rightarrow [\underline{0.9} \quad \underline{0.3}]$$

- Apple used as Technology but our static embedding tilted toward fruit.
- Need **contextual embedding** instead of static embedding.
- It should automatically increased value of .3, when sees the word **phone** and **launched** and should decrease weight of fruit

$$[\overset{\downarrow}{0.3} \quad \overset{\uparrow}{0.8}]$$

Self-Attention

- Self-attention is a mechanism that generate the smart contextual embedding from earlier generated static embedding



Transformer

- Gen-AI → center is Transformer → center is Self-Attention.
- For any NLP application we represent word into numbers
 - Word embedding → static

Money Bank



0.6	0.2	0.1	0.7
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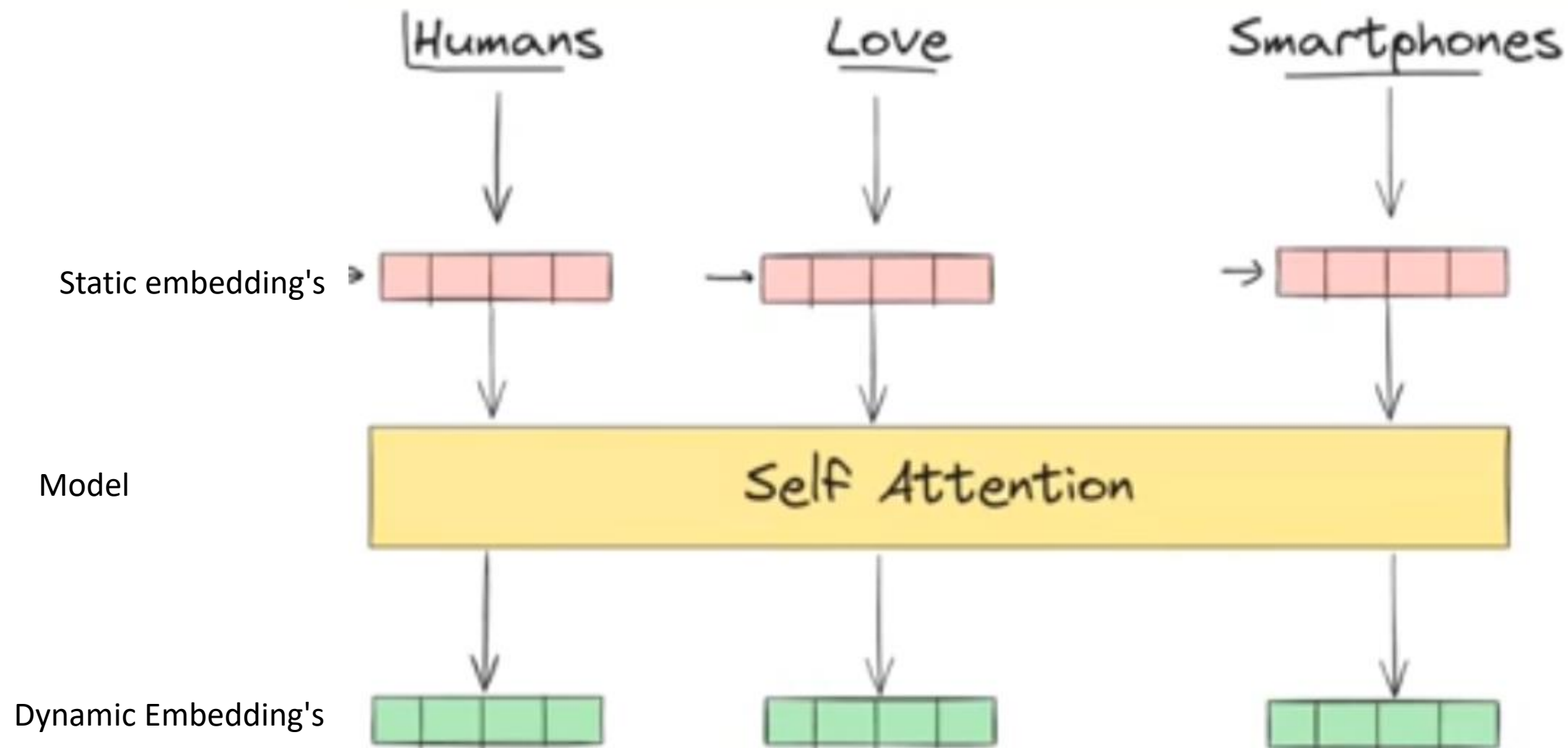
River Bank



0.6	0.2	0.1	0.7
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- Should be dynamic/contextual instead of static embedding's.

Self-Attention-Abstract Architecture



Self-attention-Box

Money Bank grows

bank \rightarrow bank

bank $\rightarrow 0.3\text{money} + 0.7\text{bank} + 0.1\text{grow}$

money = $0.7 \text{ money} + 0.2 \text{ bank} + 0.1 \text{ grows}$

bank = $0.25 \text{ money} + 0.7 \text{ bank} + 0.05 \text{ grows}$

grows = $0.1 \text{ money} + 0.2 \text{ bank} + 0.7 \text{ grows}$

River Bank Flows

bank \rightarrow bank

bank $\rightarrow 0.5\text{river} + 0.4\text{bank} + 0.1\text{flows}$

river = $0.8 \text{ river} + 0.15 \text{ bank} + 0.05 \text{ flows}$

bank = $0.2 \text{ river} + 0.78 \text{ bank} + 0.02 \text{ flows}$

flows = $0.4 \text{ river} + 0.01 \text{ bank} + 0.59 \text{ flows}$

Here words are used to represent. But machine only works with numbers

money = 0.7 money + 0.2 bank + 0.1 grows
 bank = 0.25 money + 0.7 bank + 0.05 grows
grows = 0.1 money + 0.2 bank + 0.7 grows

river = 0.8 river + 0.15 bank + 0.05 flows
 bank = 0.2 river + 0.78 bank + 0.02 flows
flows = 0.4 river + 0.01 bank + 0.59 flows

N-dimensions

N-dimensions

$$e_{\text{money}}^{(\text{new})} = 0.7 e_{\text{money}} + 0.2 e_{\text{bank}} + 0.1 e_{\text{grows}}$$

$$e_{\text{bank}}^{(\text{new})} = 0.25 e_{\text{money}} + 0.7 e_{\text{bank}} + 0.05 e_{\text{grows}}$$

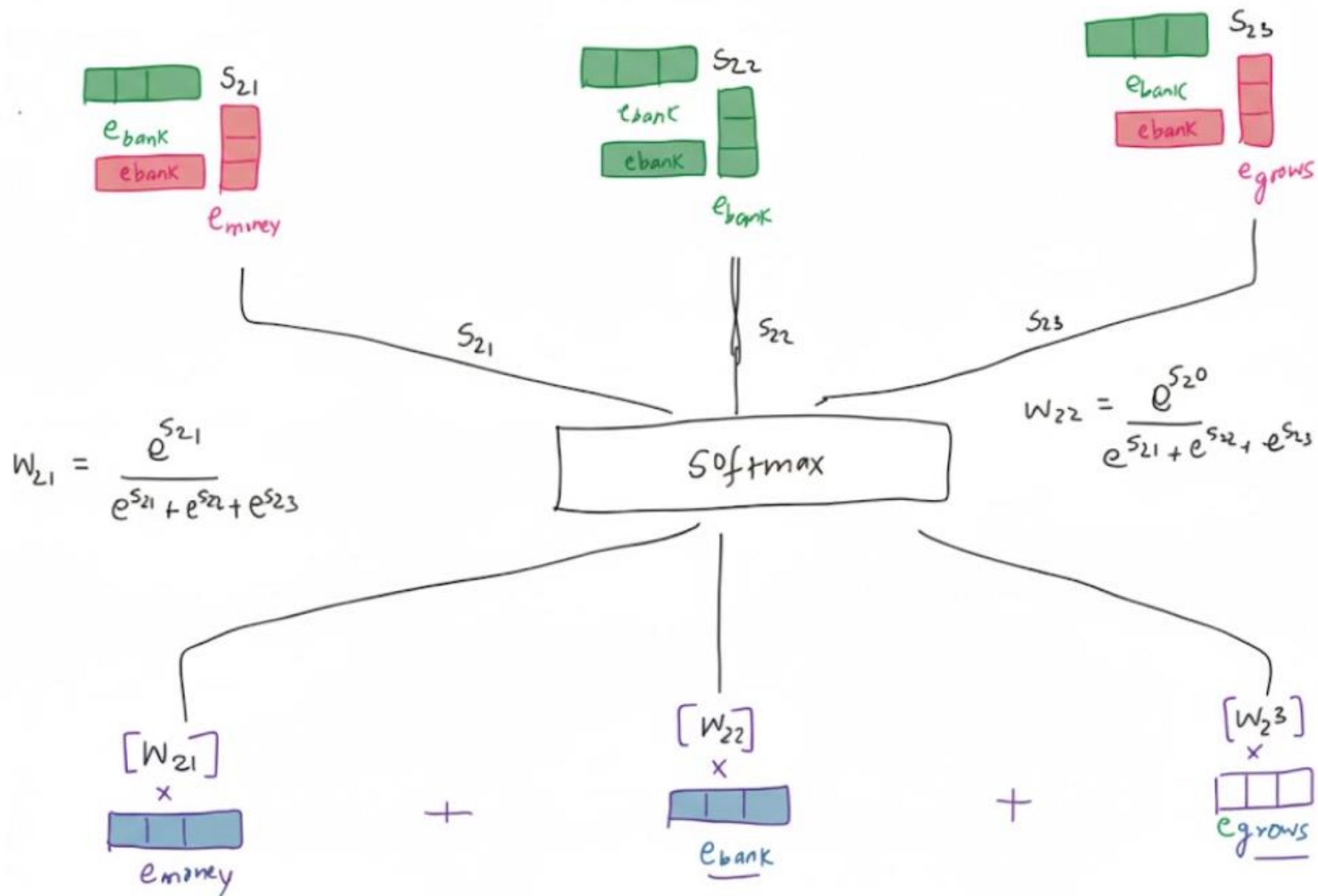
$$e_{\text{gross}}^{(\text{new})} = 0.1 e_{\text{money}} + 0.2 e_{\text{bank}} + 0.7 e_{\text{gross}}$$

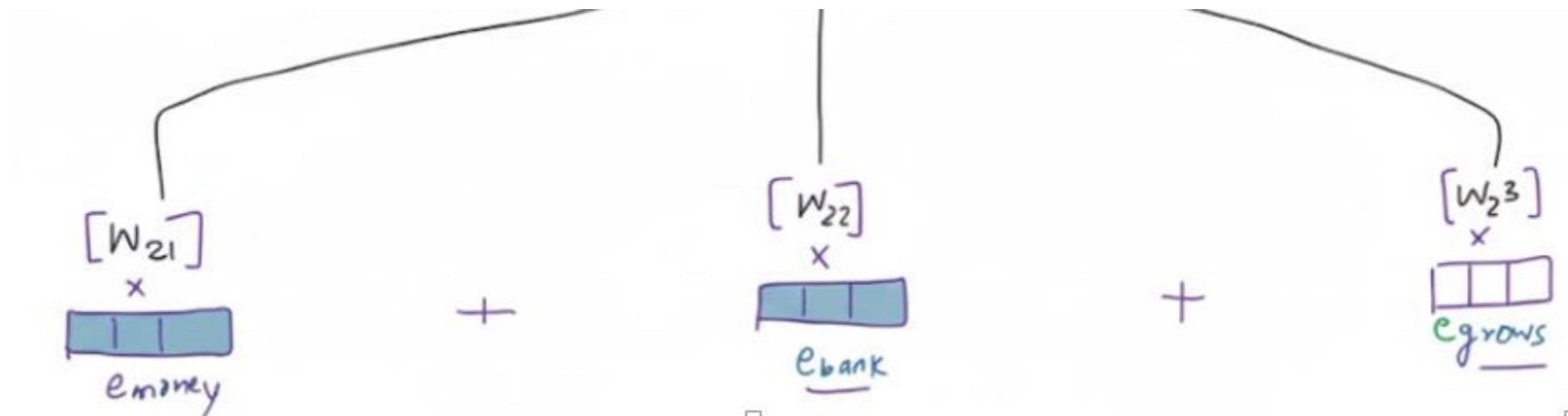
Numbers represent the similarity between the words

Similarity between the two vector can be calculated with dot product of two vectors.

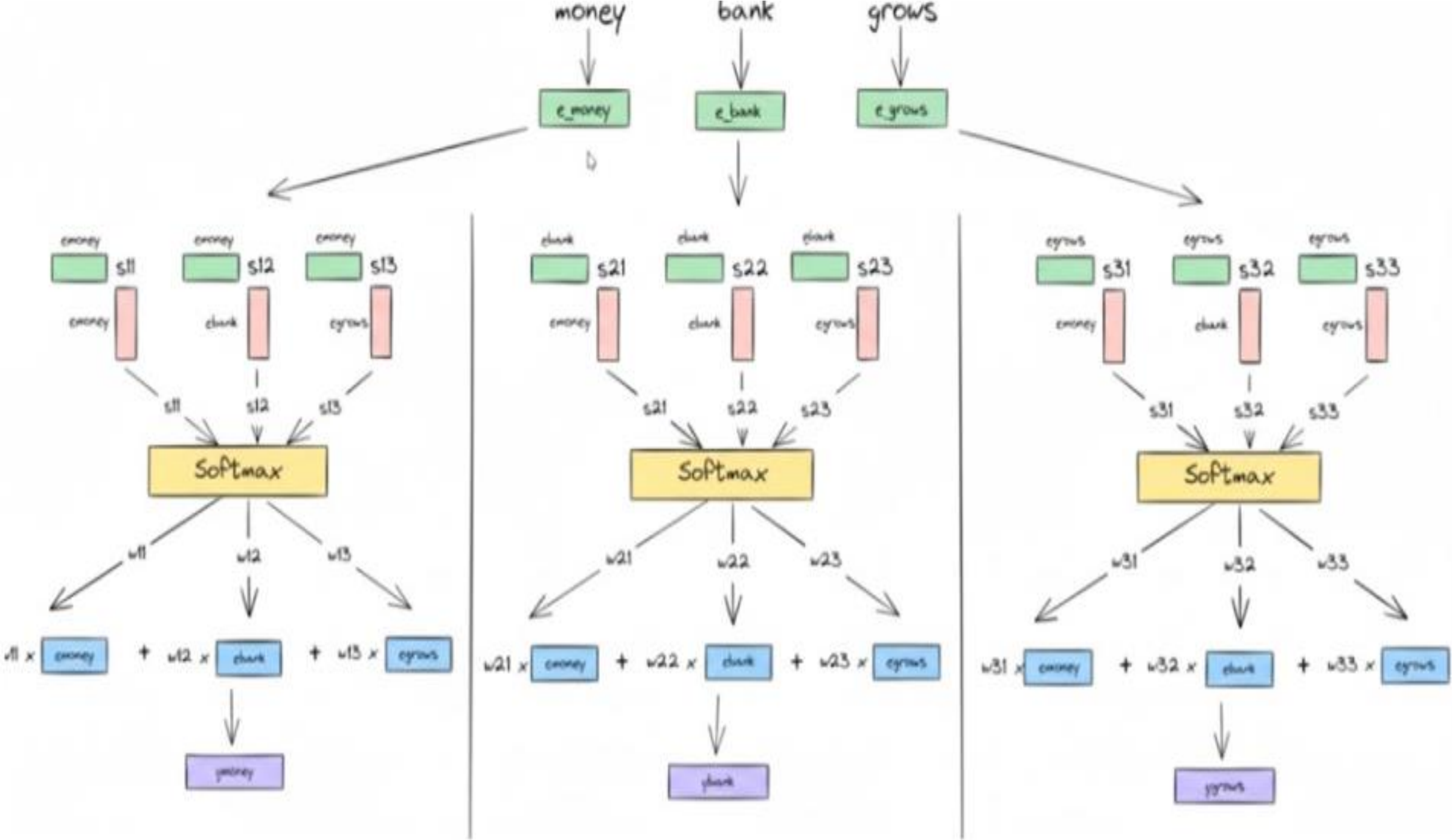
$$e_{\text{bank}}^{(\text{new})} = (e_{\text{bank}} \cdot e_{\text{money}}^T) e_{\text{money}} + (e_{\text{bank}} \cdot e_{\text{bank}}^T) e_{\text{bank}} + (e_{\text{bank}} \cdot e_{\text{grows}}^T) e_{\text{grows}}$$

$$e_{bank}^{(new)} = (e_{bank} \cdot e_{money}^T) e_{money} + (e_{bank} \cdot e_{bank}^T) e_{bank} + (e_{bank} \cdot e_{grows}^T) e_{grows}$$

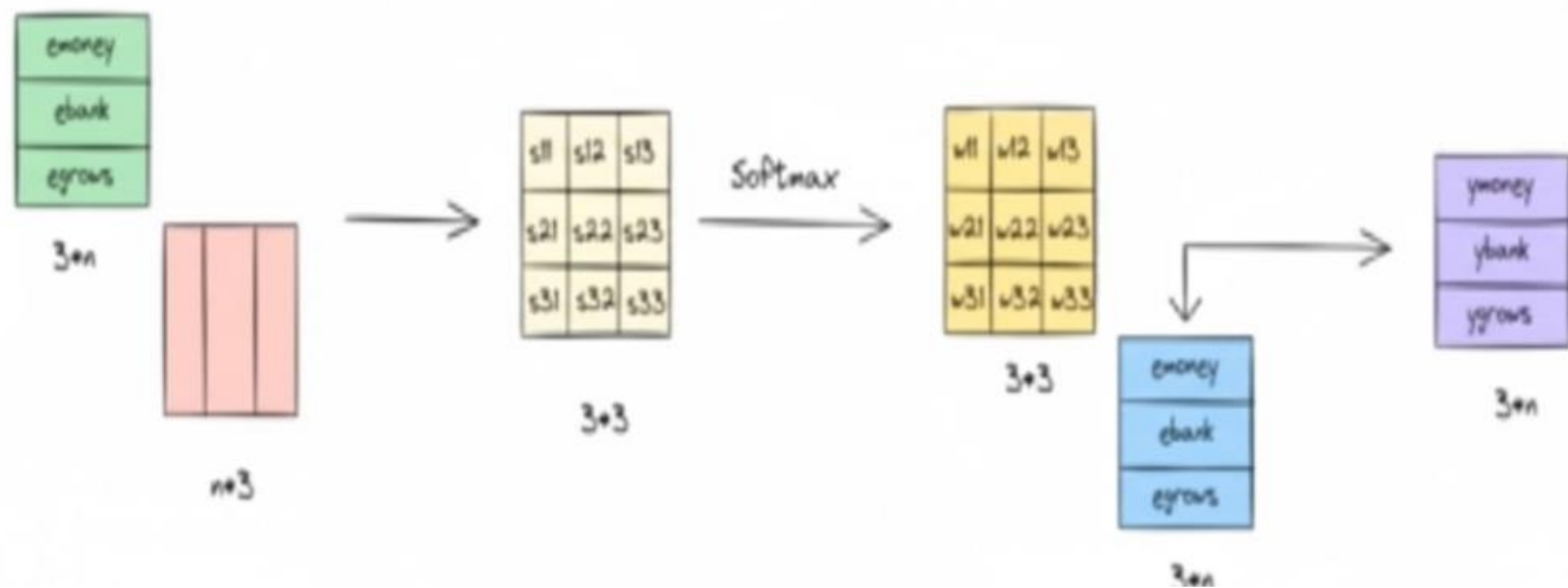




$e_{\text{bank}}^{(\text{new})} = Y_{\text{bank}} \rightarrow \text{Contextual word Embedding}$

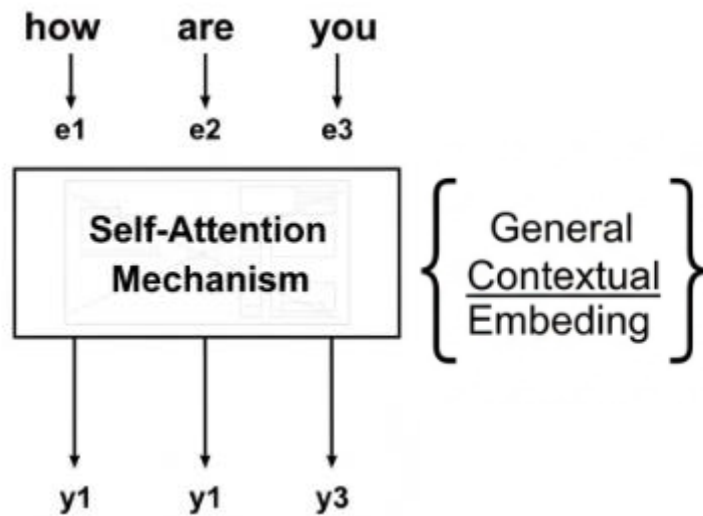


- Points to Consider
- → This operation is a parallel operation
- There are no parameters involved



There is no parameter involved

- $Hello \rightarrow e_1 \rightarrow Y_1$

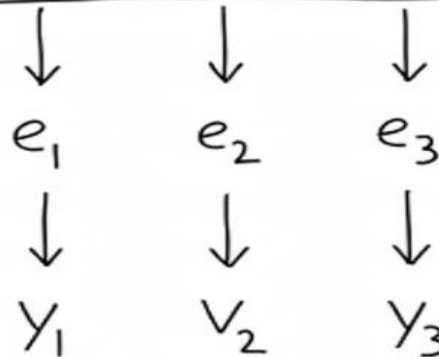


English	Urdu
How are you	کیسے ہو؟
I am good	میں ٹھیک ہوں
Piece of cake	بہت آسان کام

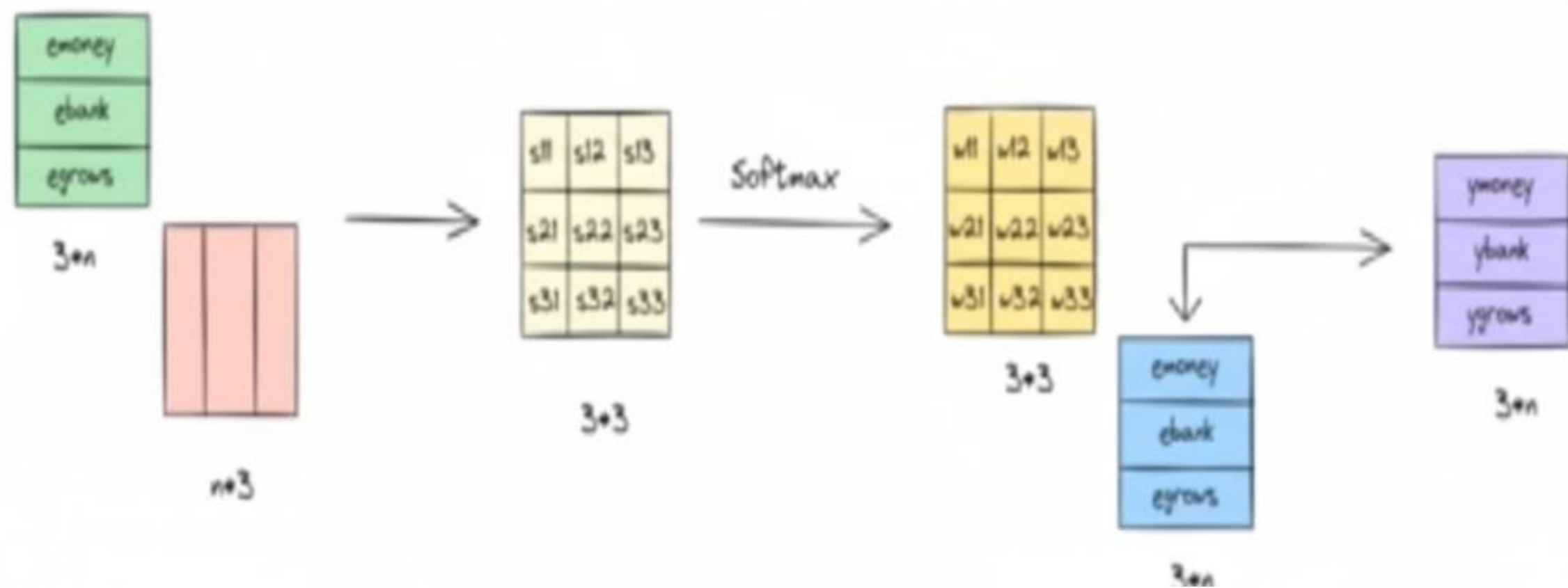
Not-Task
Specific
contextual
Embedding

piece of cake

کیک کا ٹکڑا



break a leg کامیابی ہو



Words \rightarrow machine

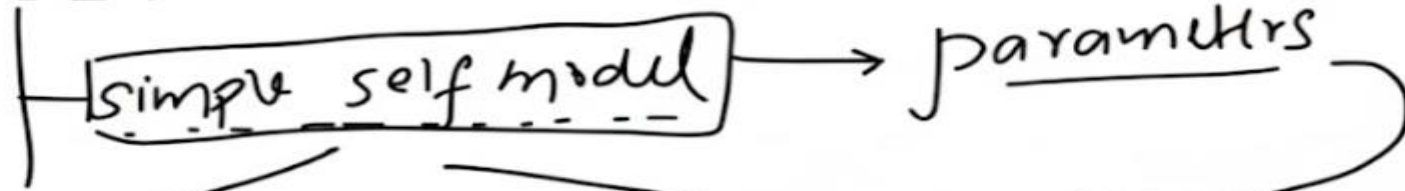


embedding \rightarrow semantic

\rightarrow Context

river bank money bank

\rightarrow contextual embeddings

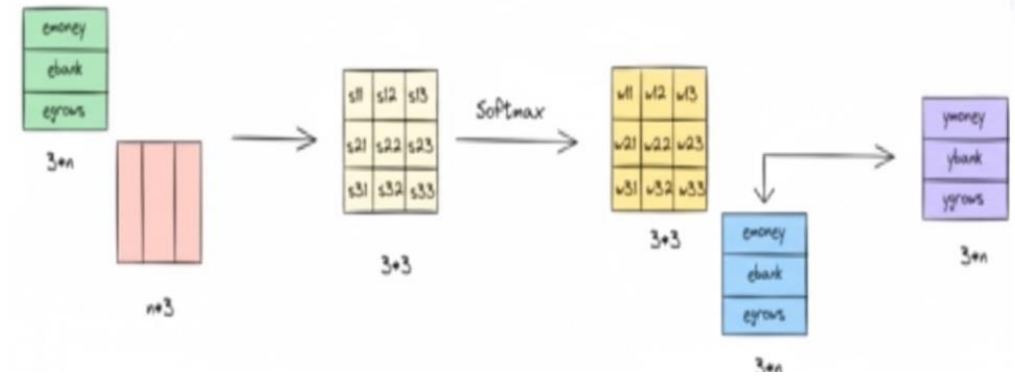


variable parameters \rightarrow general

task-specific

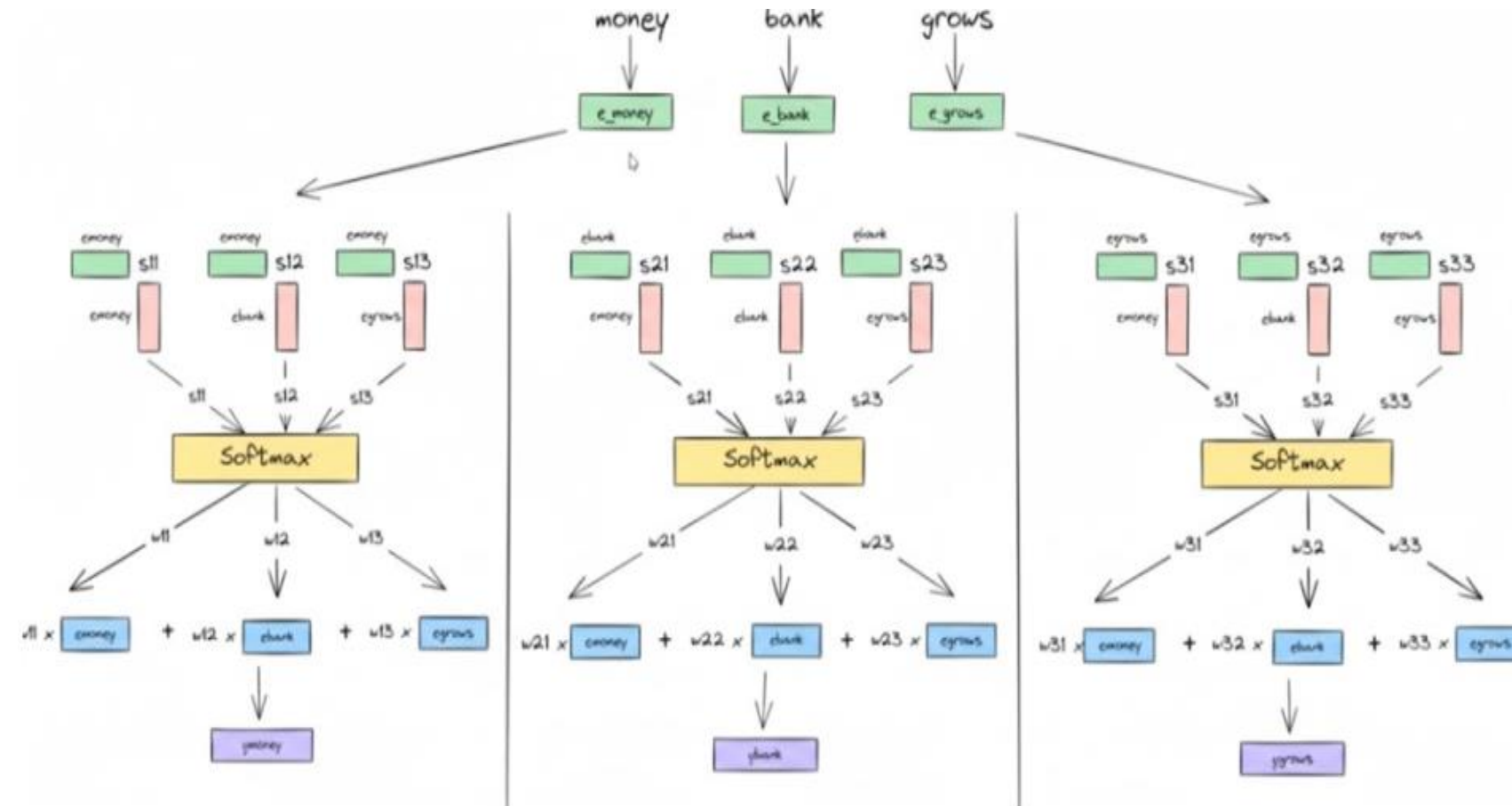
To make it Task specific , we have to add some learnable parameters

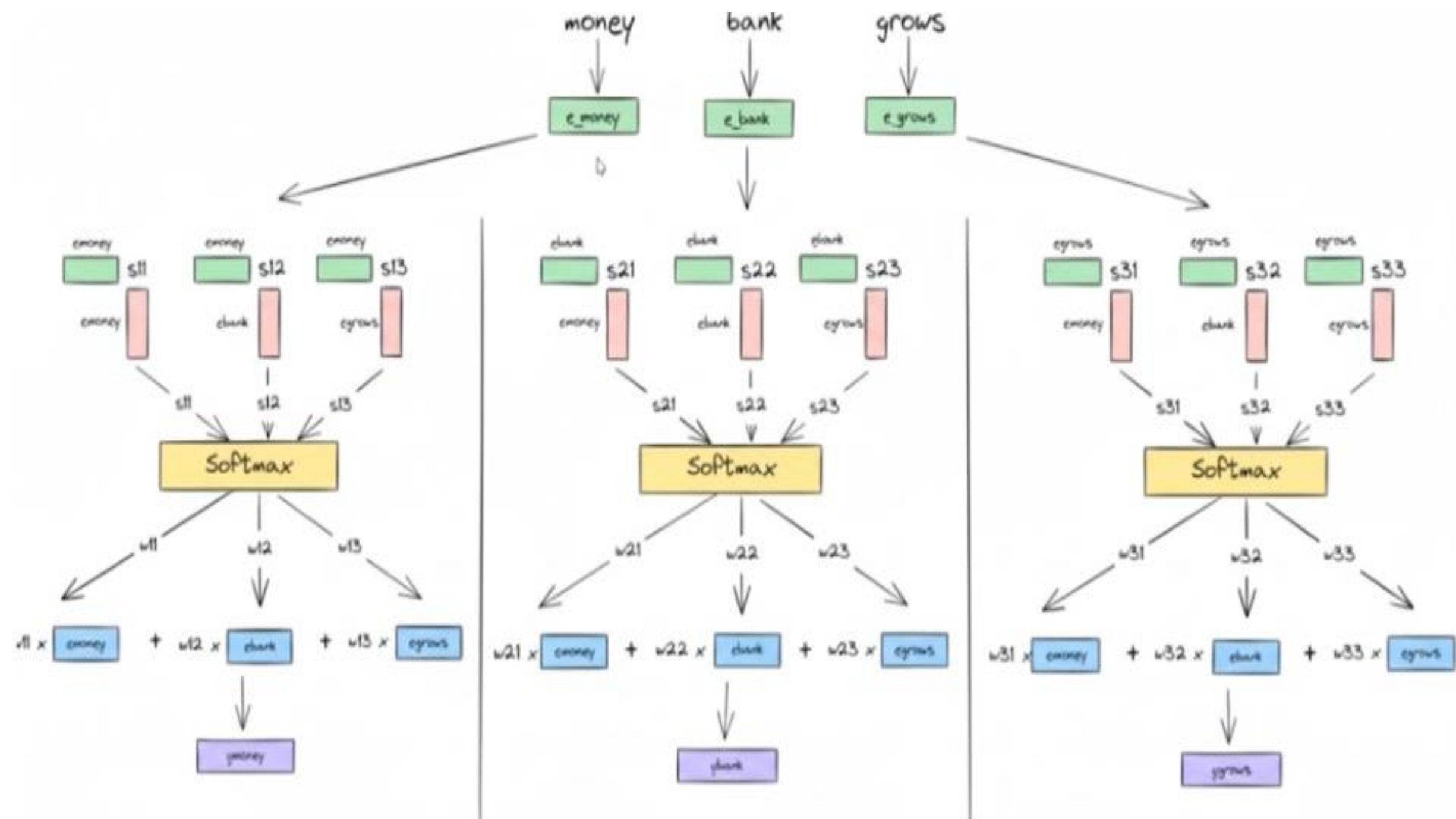
- Where will you add parameter

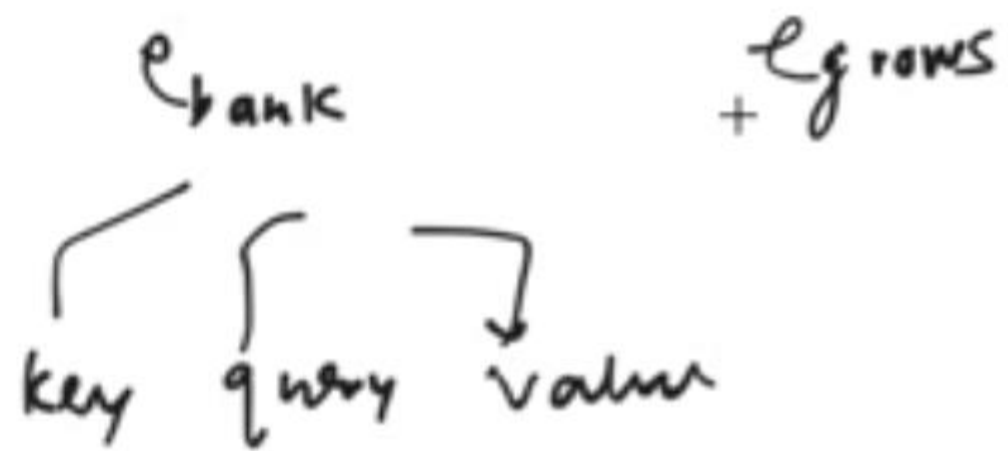
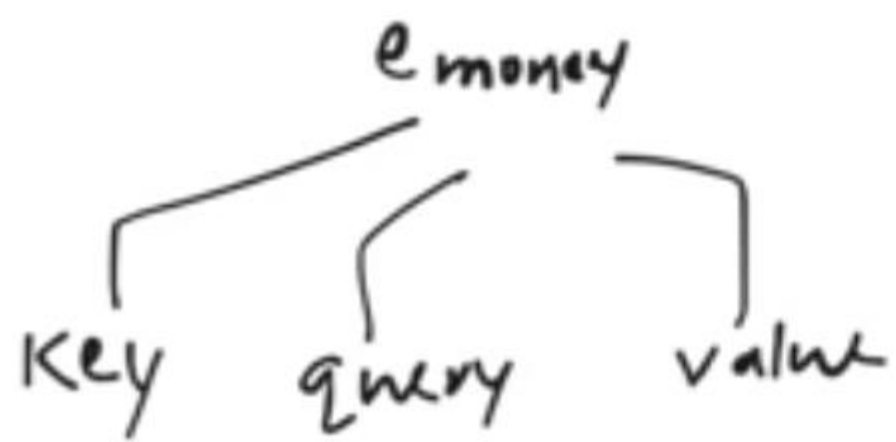


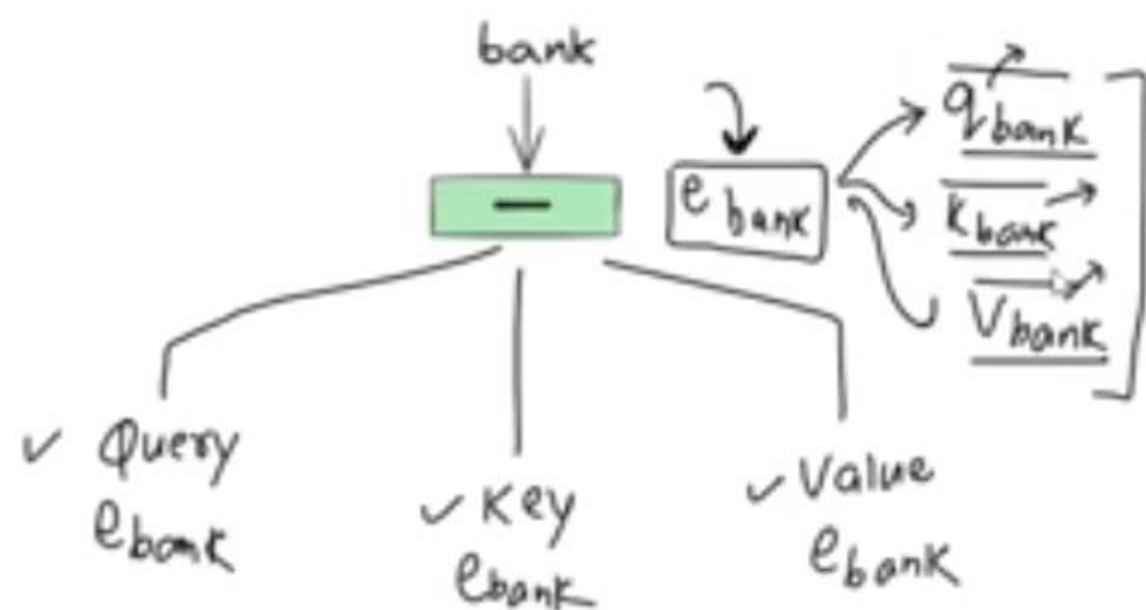
Query, Key , value

$D=[a:2, b: 3, c:3]$

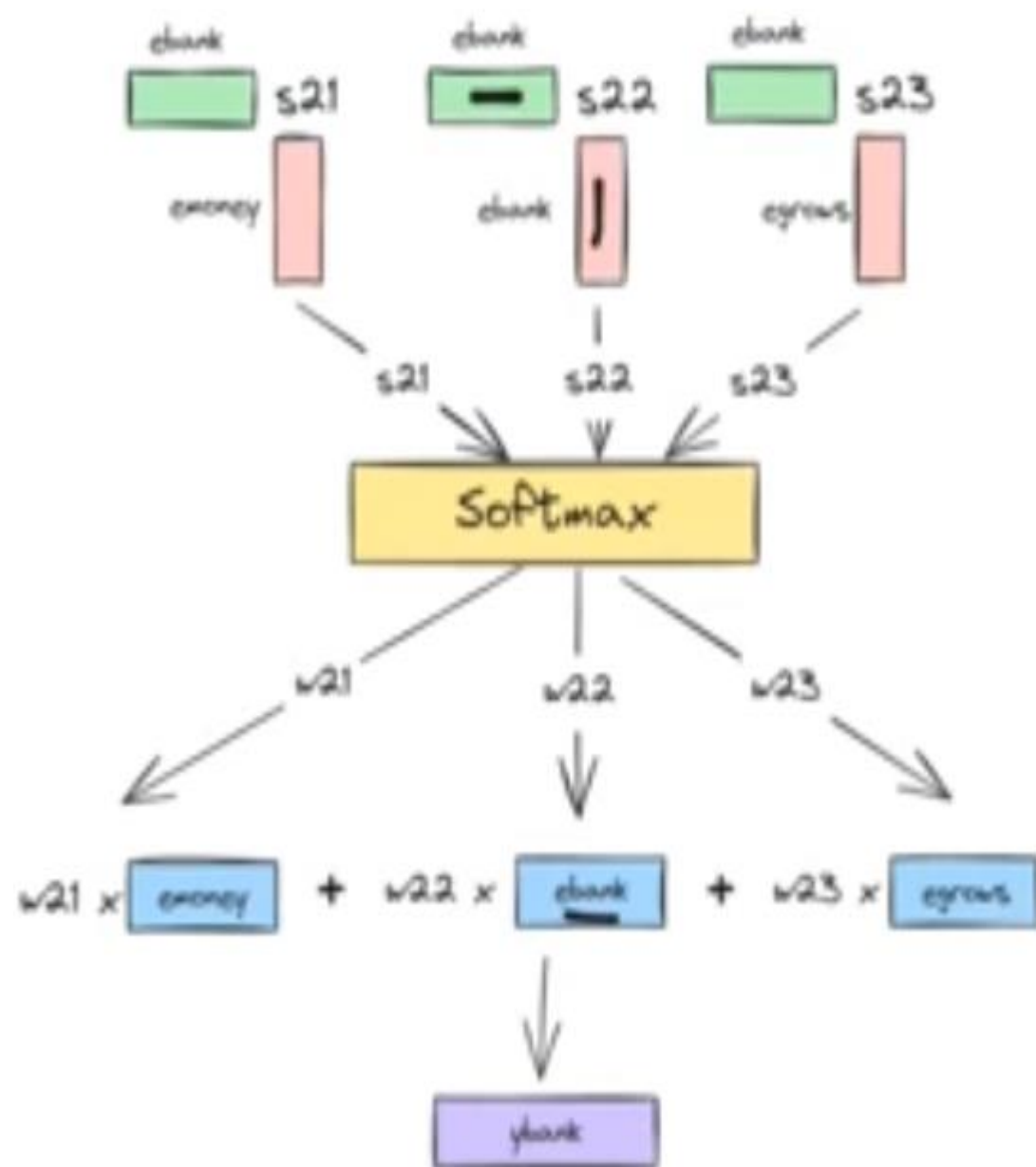








Separation of concerns



Person: Author/writer of books → autobiography
Age=35
Shadi.com

profile
specification/requirements
search(query)
match

meeting/call

Ebank

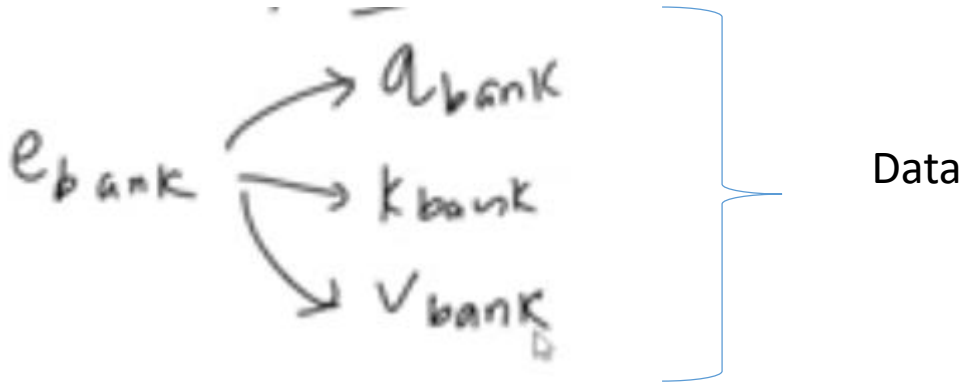
key

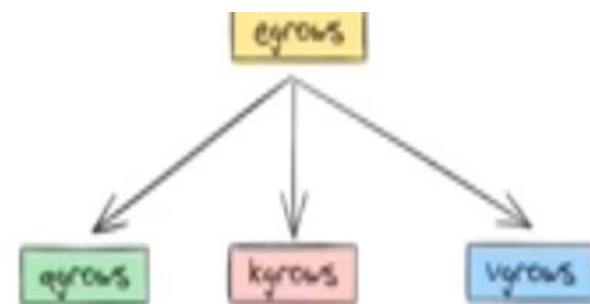
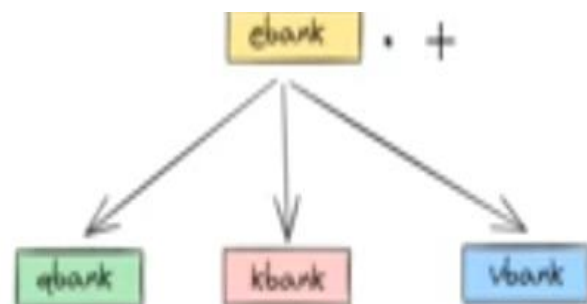
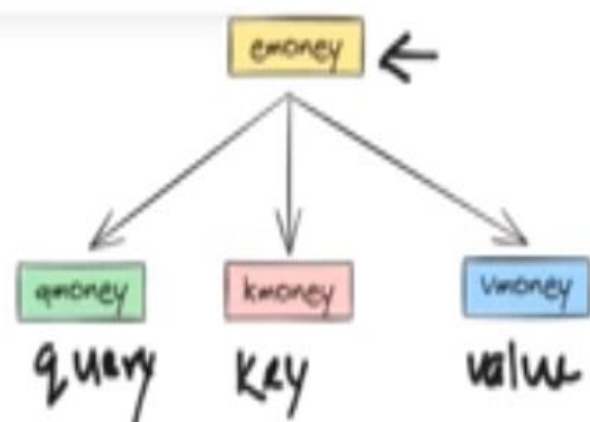
Query

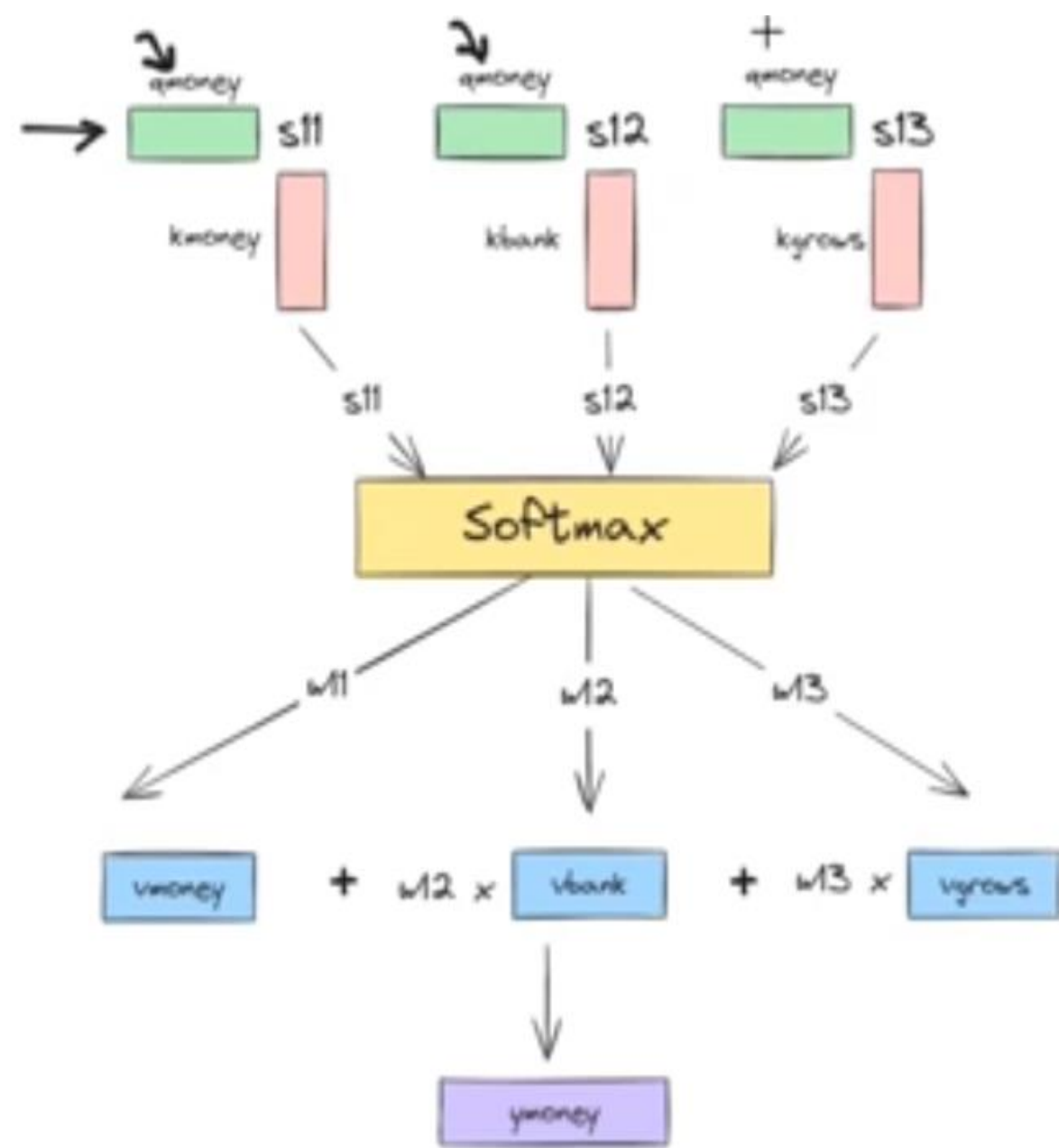
value

Autobiography

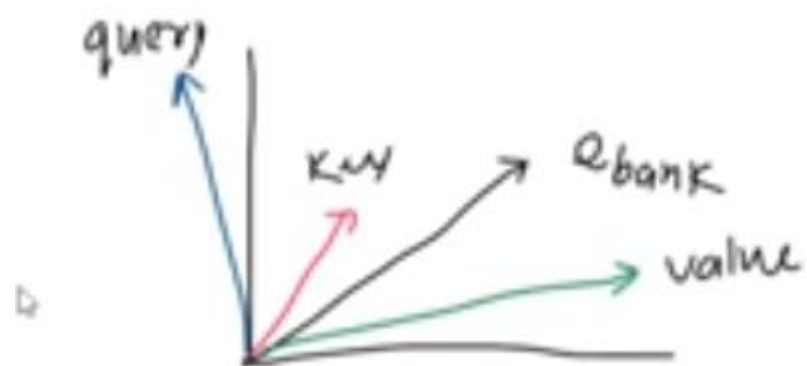
profile
search
match

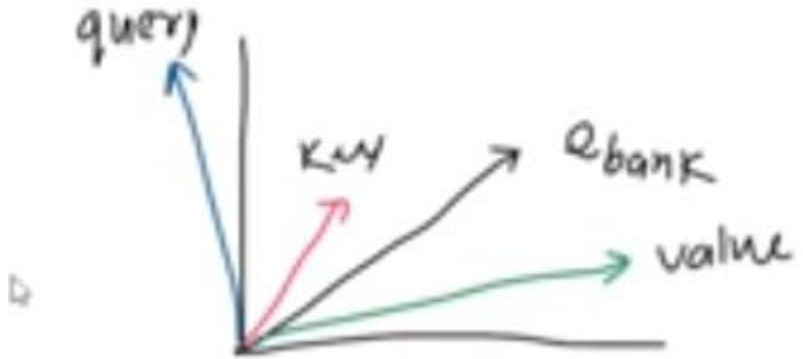






money \rightarrow





Change magnitude (increase or decrease)
 Linear Transformation (change direction)

Training

Forward propagation
 backward propagation

