

Self-Attention

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- For any NLP task , we give the input a set of words → 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

- Problem:

DHE $\begin{bmatrix} 1 & \boxed{\text{mat}} & \boxed{\text{cat}} & \boxed{\text{rat}} \\ 0 & \text{cat} & \text{rat} & \text{rat} \end{bmatrix}$

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

king $\rightarrow [0.6 \ 0.1 \ 1 \ 0 \ 0.9]$

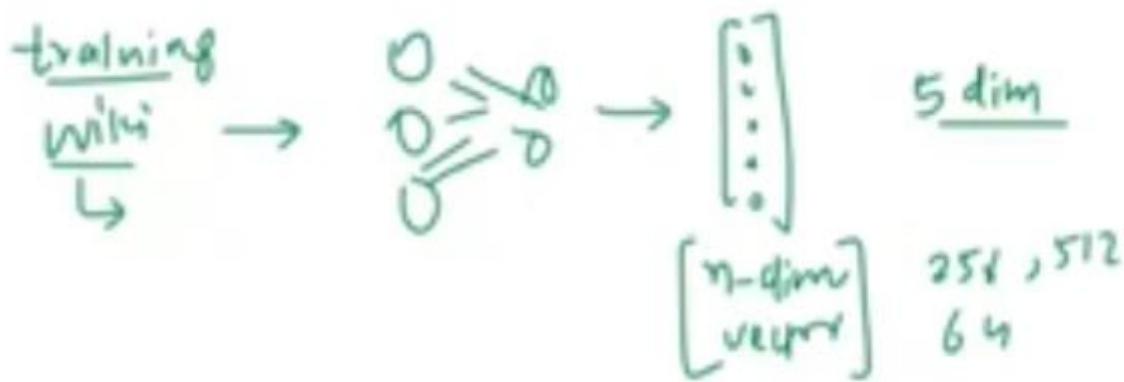
queen $\rightarrow [0.3 \ 0.2 \ 0.4 \ 1 \ 0]$

- BoW(how many time a word came)

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

- Tf-Idf

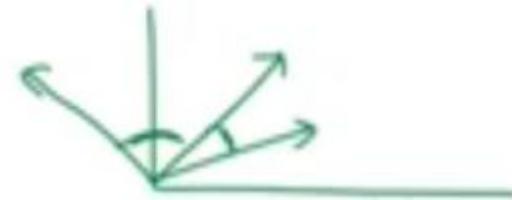
- Word Embedding (Most Powerful having Semantics)



- 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 $\rightarrow [0.3 \ 0.2 \ 0.4 \ 1 \ 0]$

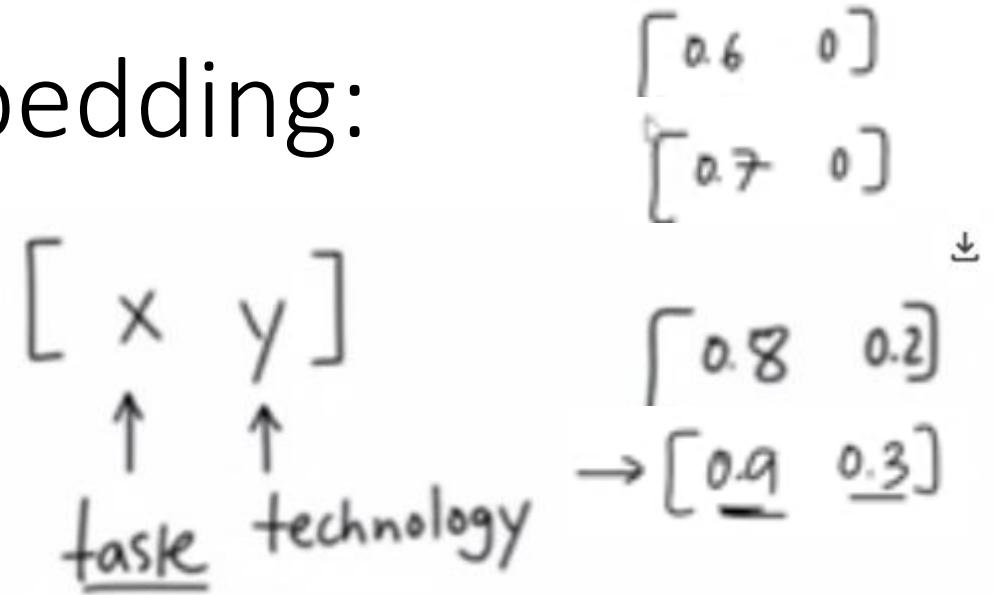
citizen $\{0.2 \ 0.9 \ 0.9 \ 0.9\}$



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.

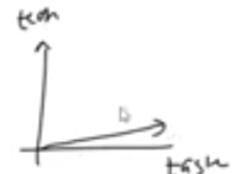
- Word Embedding → Static Make once and used again& again

- Make once and used again &again

- It should be dynamic

Fruit → [Apple Launched a new Phone
while I was eating an orange]

static → [0.9 0.3]



- Self-attention produce dynamic embedding from static embedding

- Static embedding is:

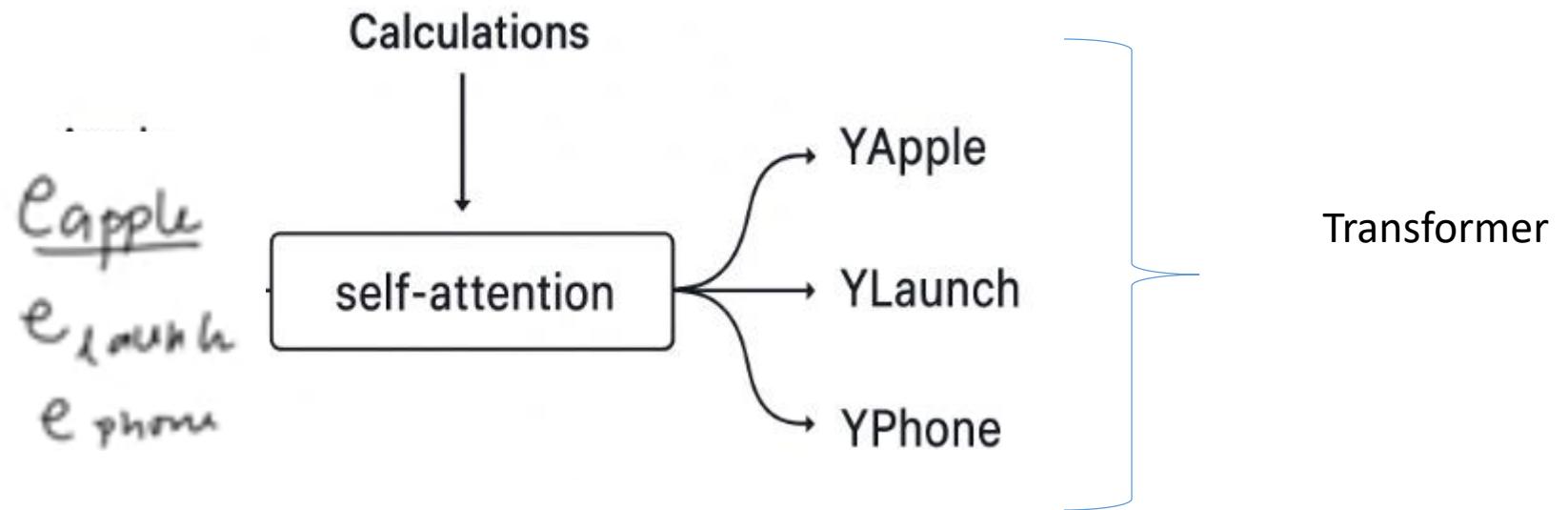
$$\rightarrow [\underline{0.9} \quad \underline{0.3}]$$

- 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

$$\begin{bmatrix} \downarrow & \\ [0.3 & 0.8] \end{bmatrix}$$

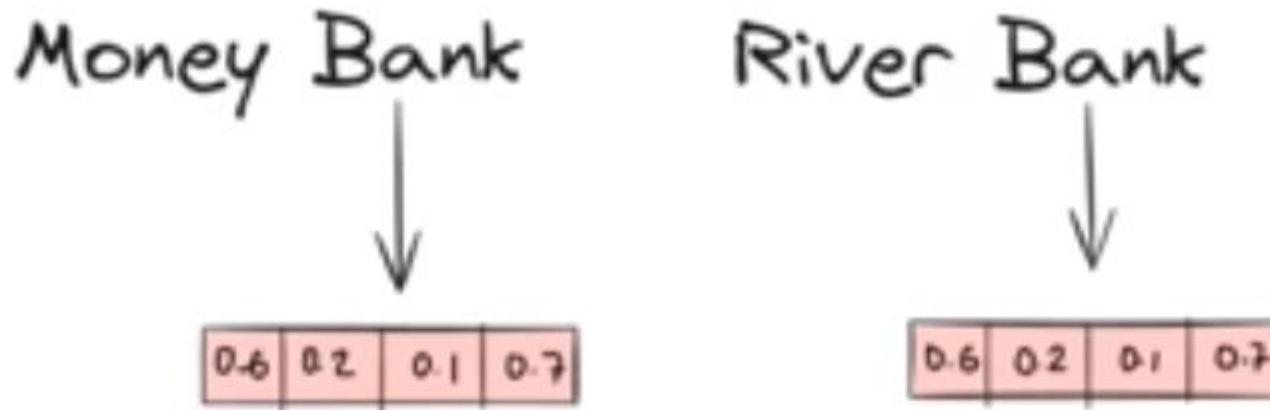
Self-Attention

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



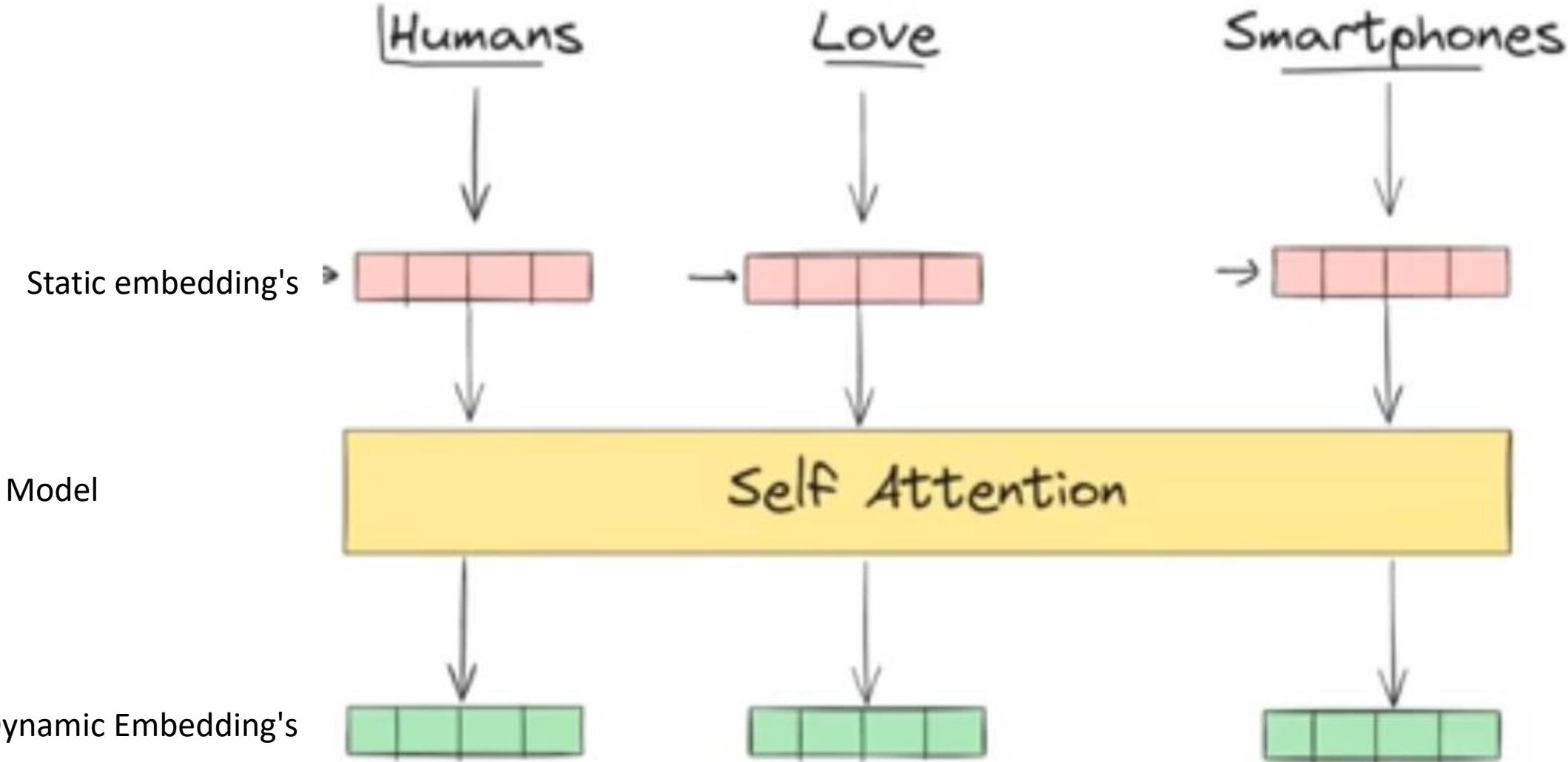
Transformer

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



- Should be dynamic/contextual instead of static embedding's.

Self-Attention-Abstract Architecture



Self-attention-Box

Money Bank grows

bank → bank

bank → 0.3money + 0.7bank + 0.1grow

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 Bank Flows

bank → bank

bank → 0.5river + 0.4bank + 0.1flows

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

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

$$\begin{aligned} \text{money} &= 0.7 \text{ money} + 0.2 \text{ bank} + 0.1 \text{ grows} \\ \text{bank} &= 0.25 \text{ money} + 0.7 \text{ bank} + 0.05 \text{ grows} \\ \text{grows} &= 0.1 \text{ money} + 0.2 \text{ bank} + 0.7 \text{ grows} \end{aligned}$$

$$\begin{aligned} \text{river} &= 0.8 \text{ river} + 0.15 \text{ bank} + 0.05 \text{ flows} \\ \text{bank} &= 0.2 \text{ river} + 0.78 \text{ bank} + 0.02 \text{ flows} \\ \text{flows} &= 0.4 \text{ river} + 0.01 \text{ bank} + 0.59 \text{ flows} \end{aligned}$$

N-dimensions

N-dimensions

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

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

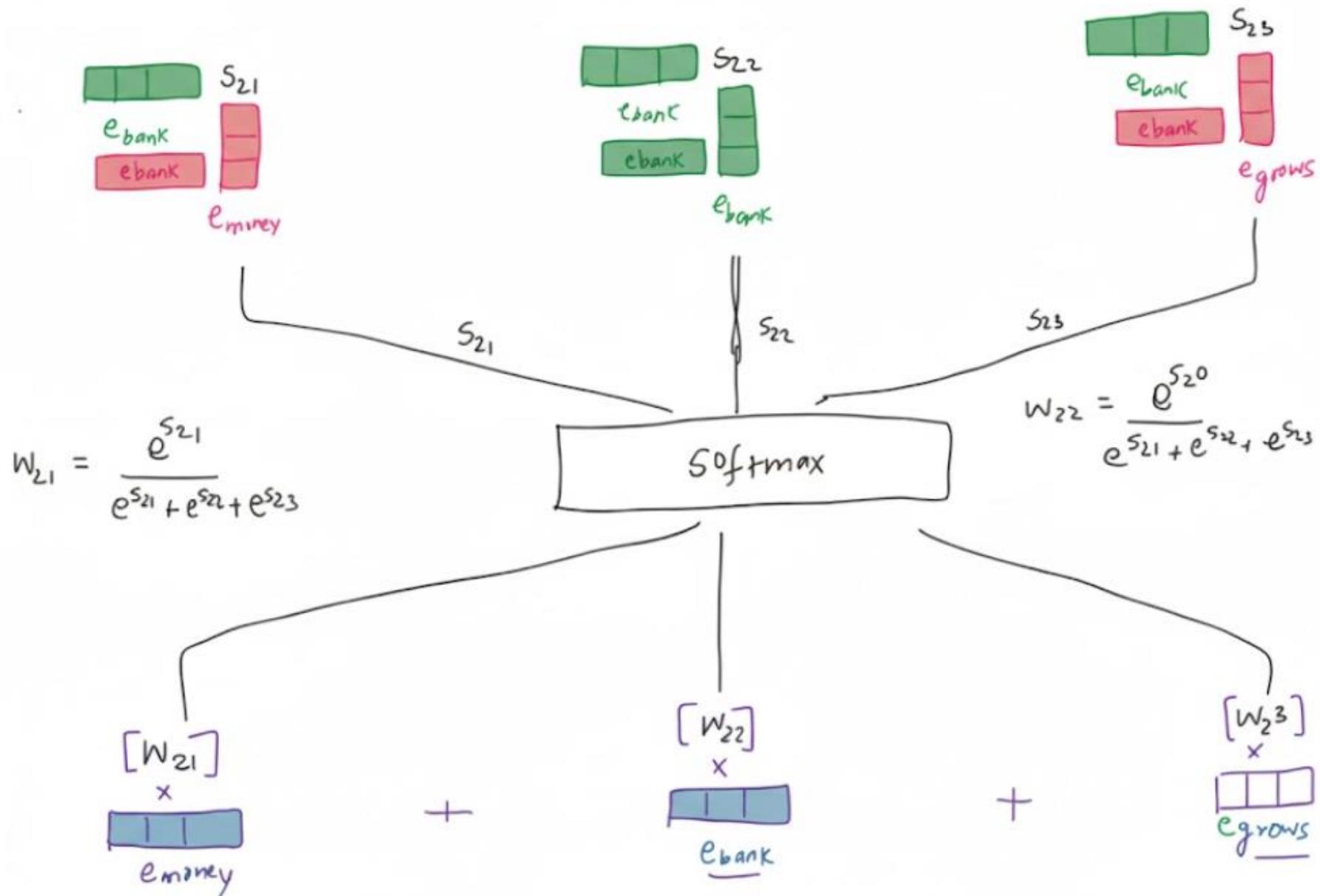
$$e_{\text{gross}}^{(\text{new})} = 0.1 e_{\text{money}} + 0.2 e_{\text{bank}} + 0.7 e_{\text{gross}}$$

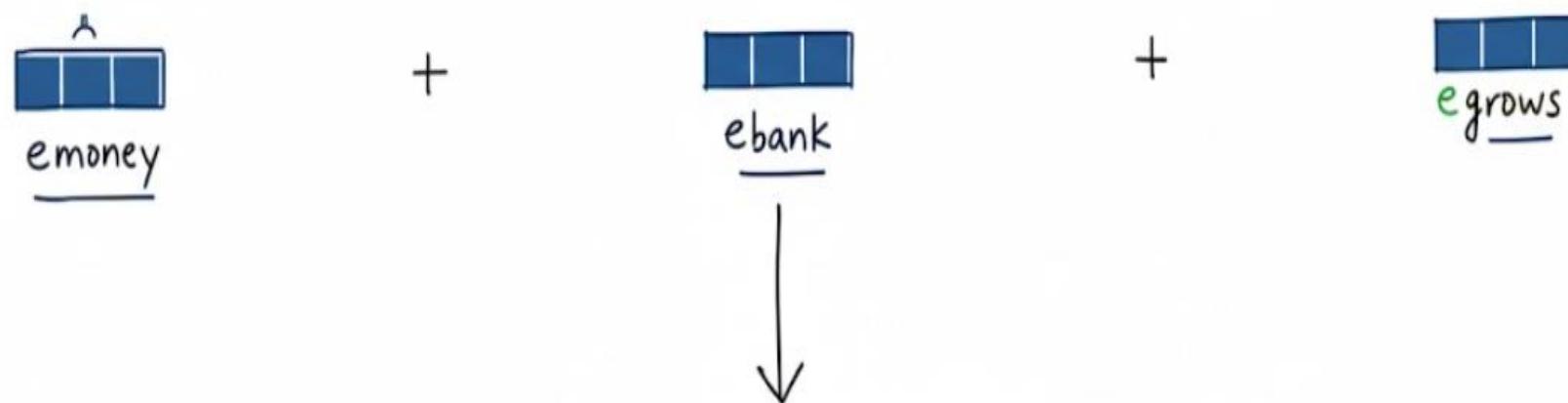
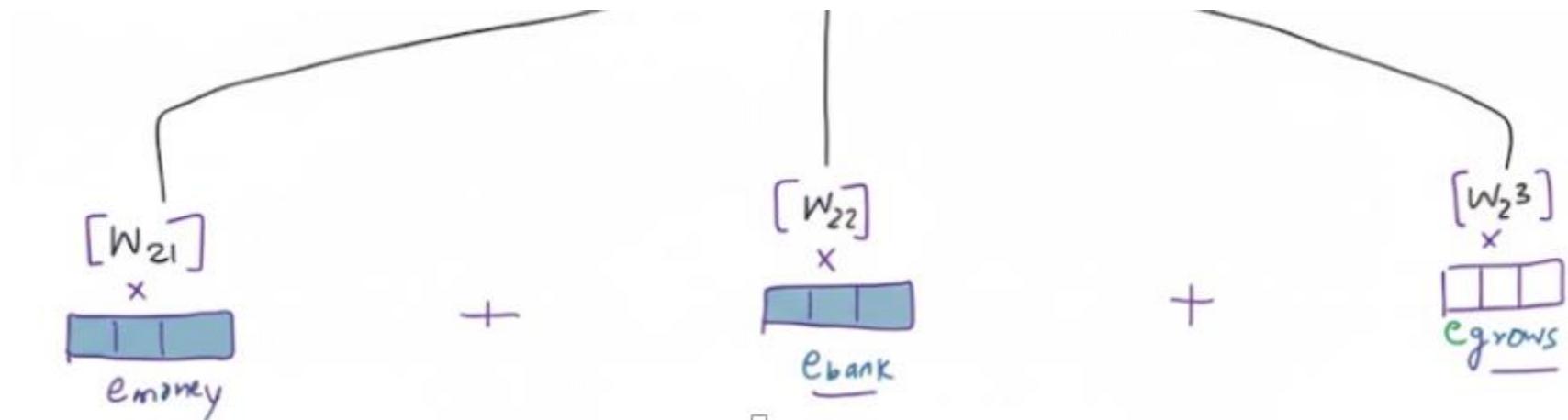
Numbers represent the similarity between the words

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

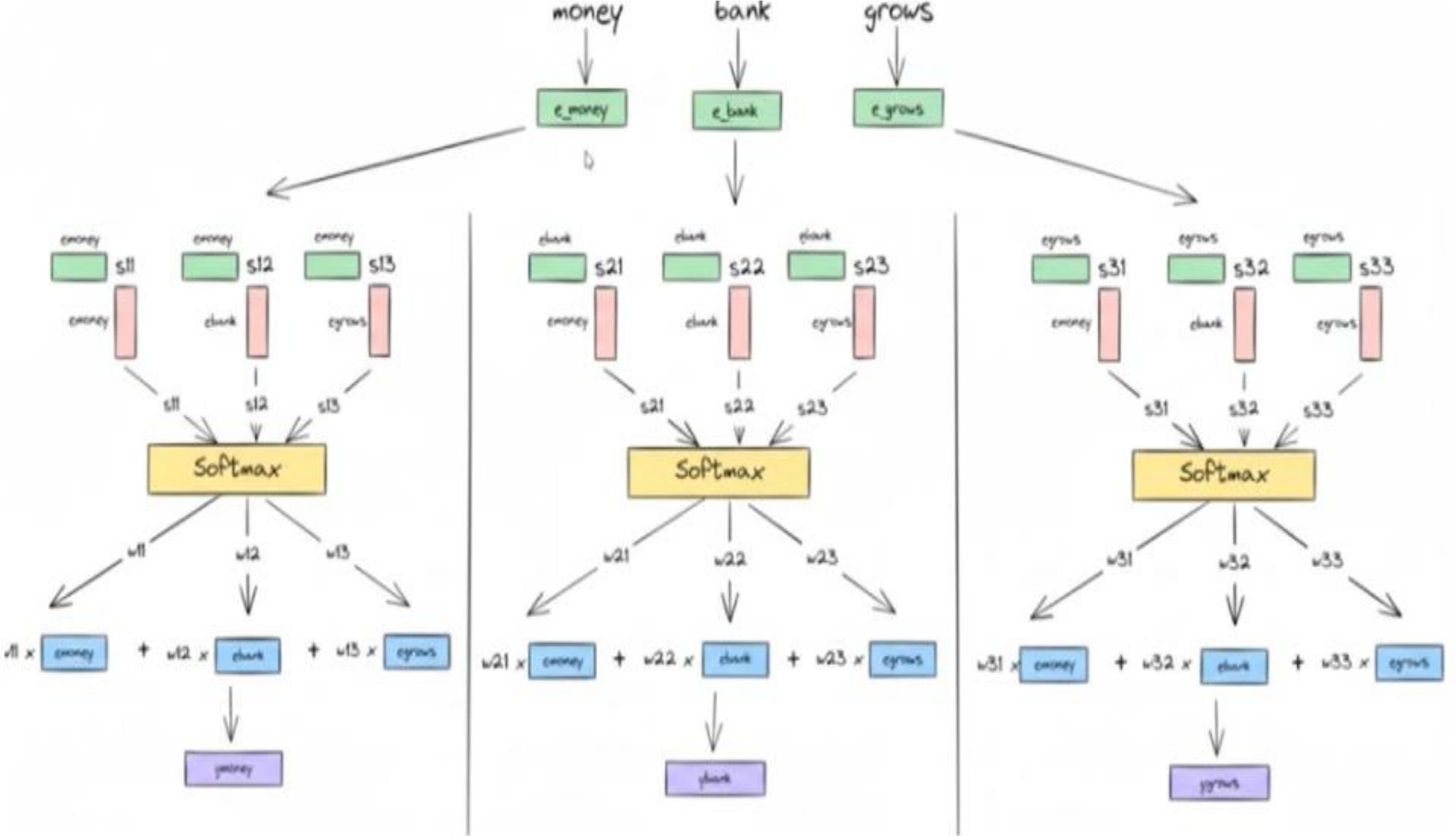
$$e_{\text{bank}}^{(\text{new})} = (e_{\text{bank}} \cdot e_{\text{money}}^T) e_{\text{money}} + (e_{\text{bank}} \cdot e_{\text{bank}}^T) e_{\text{bank}} + (e_{\text{bank}} \cdot e_{\text{grows}}^T) e_{\text{grows}}$$

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

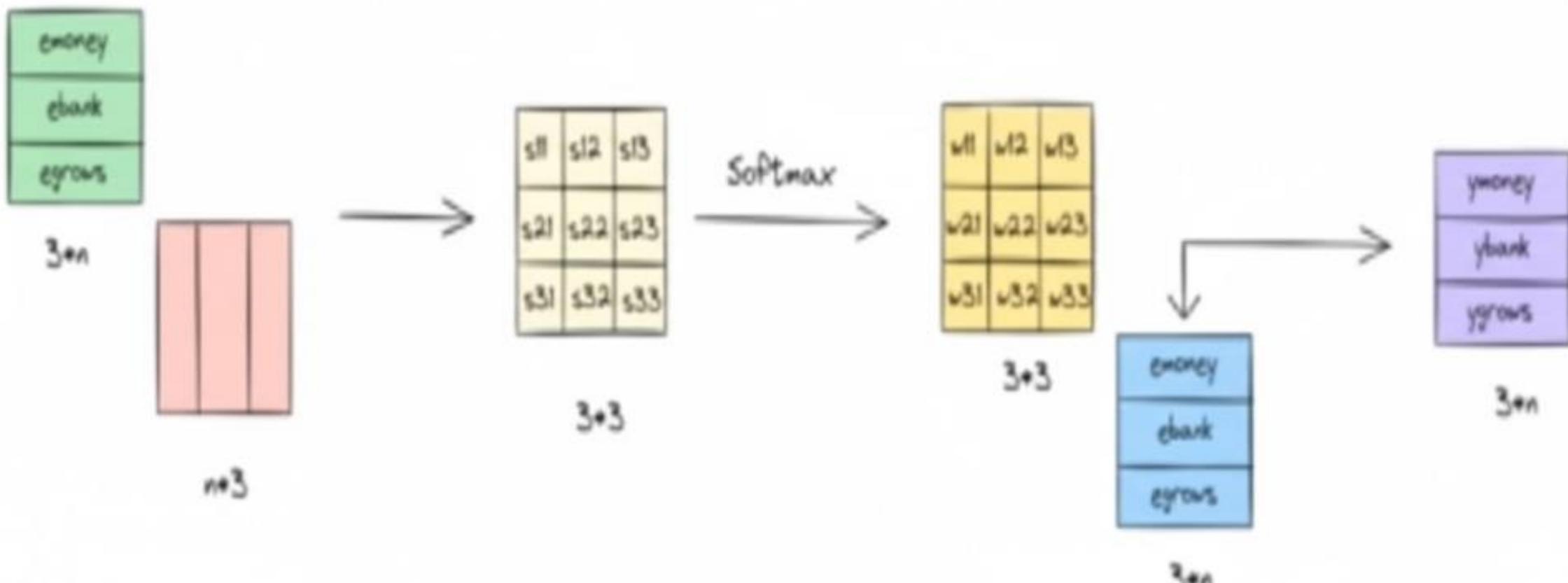




$$e_{\text{bank}}^{(\text{new})} = Y_{\text{bank}} \rightarrow \text{Contextual word Embedding}$$

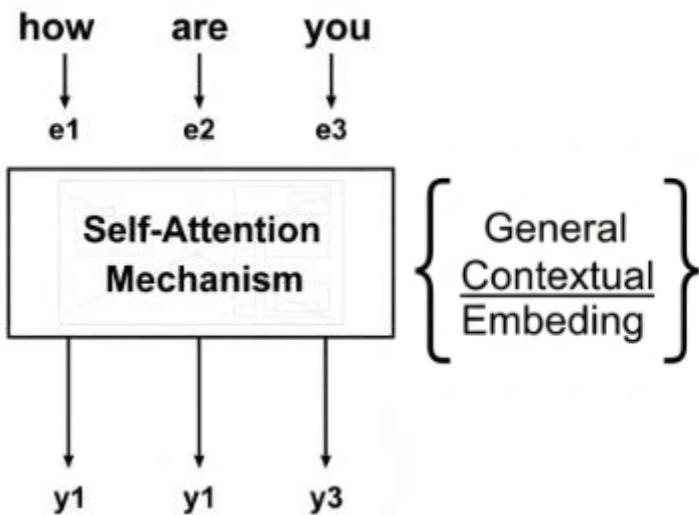


- Points to Consider
- → This operation is a parallel operation
 - There are no parameters involved

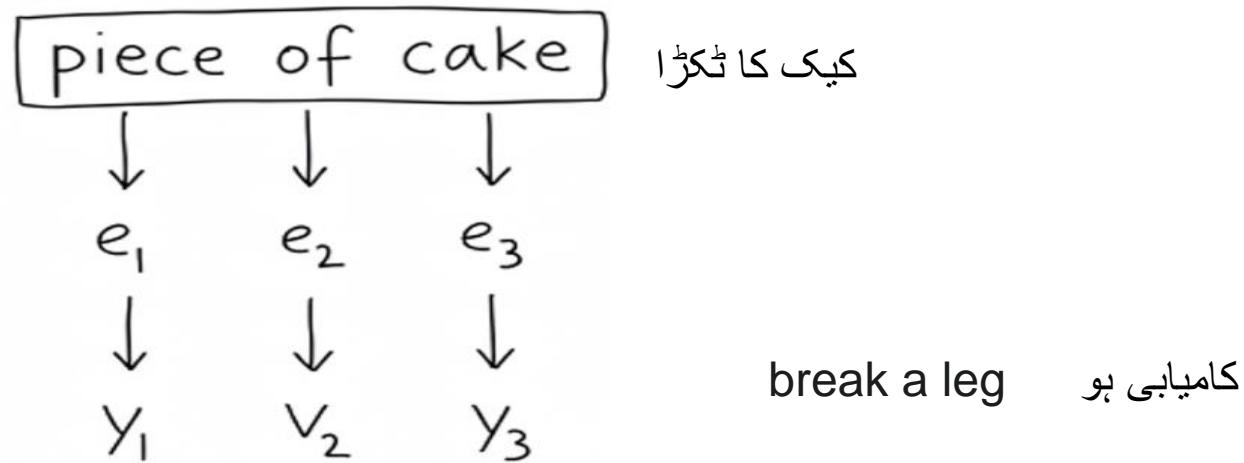


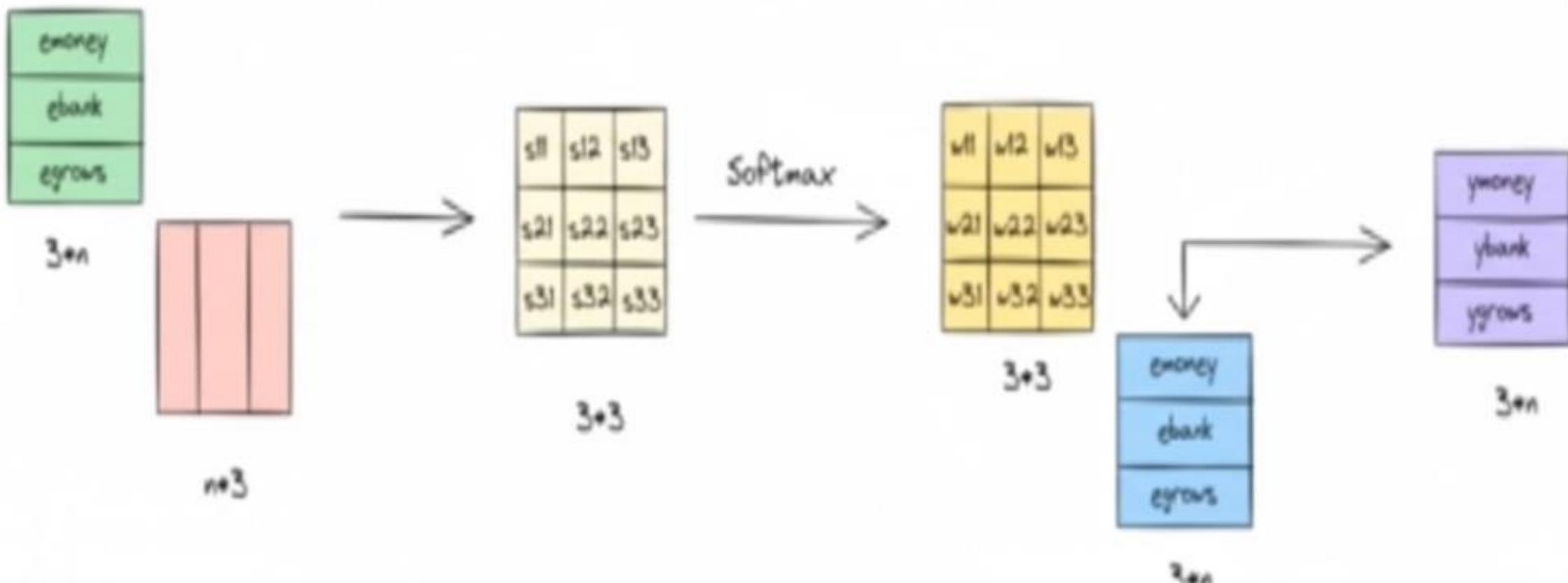
There is no parameter involved

- $Hello \rightarrow e_1 \rightarrow Y_1$



English	Urdu	Not-Task Specific contextual Embedding
How are you	کیسے ہو؟	
I am good	میں ٹھیک ہوں	
Piece of cake	بہت آسان کام	

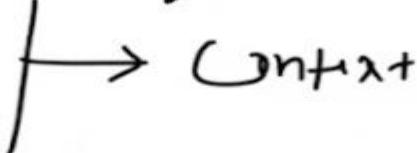




words → machine



embedding → semantic



river bant money bank

→ contextual embeddings

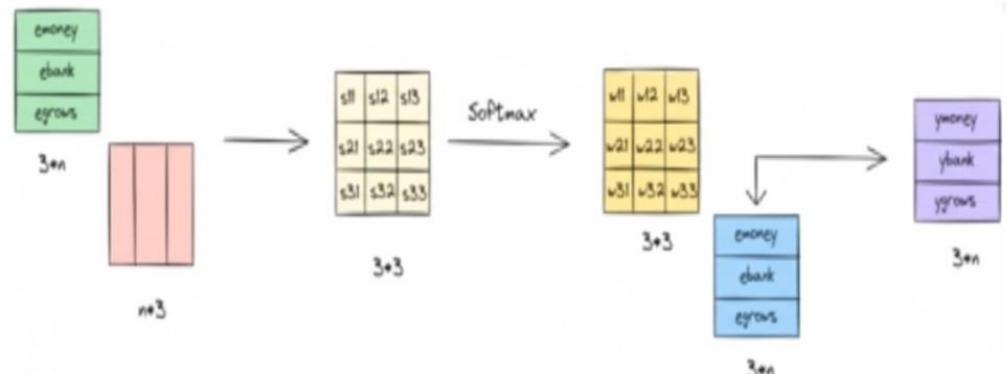
↓
|
simple self model

parameters

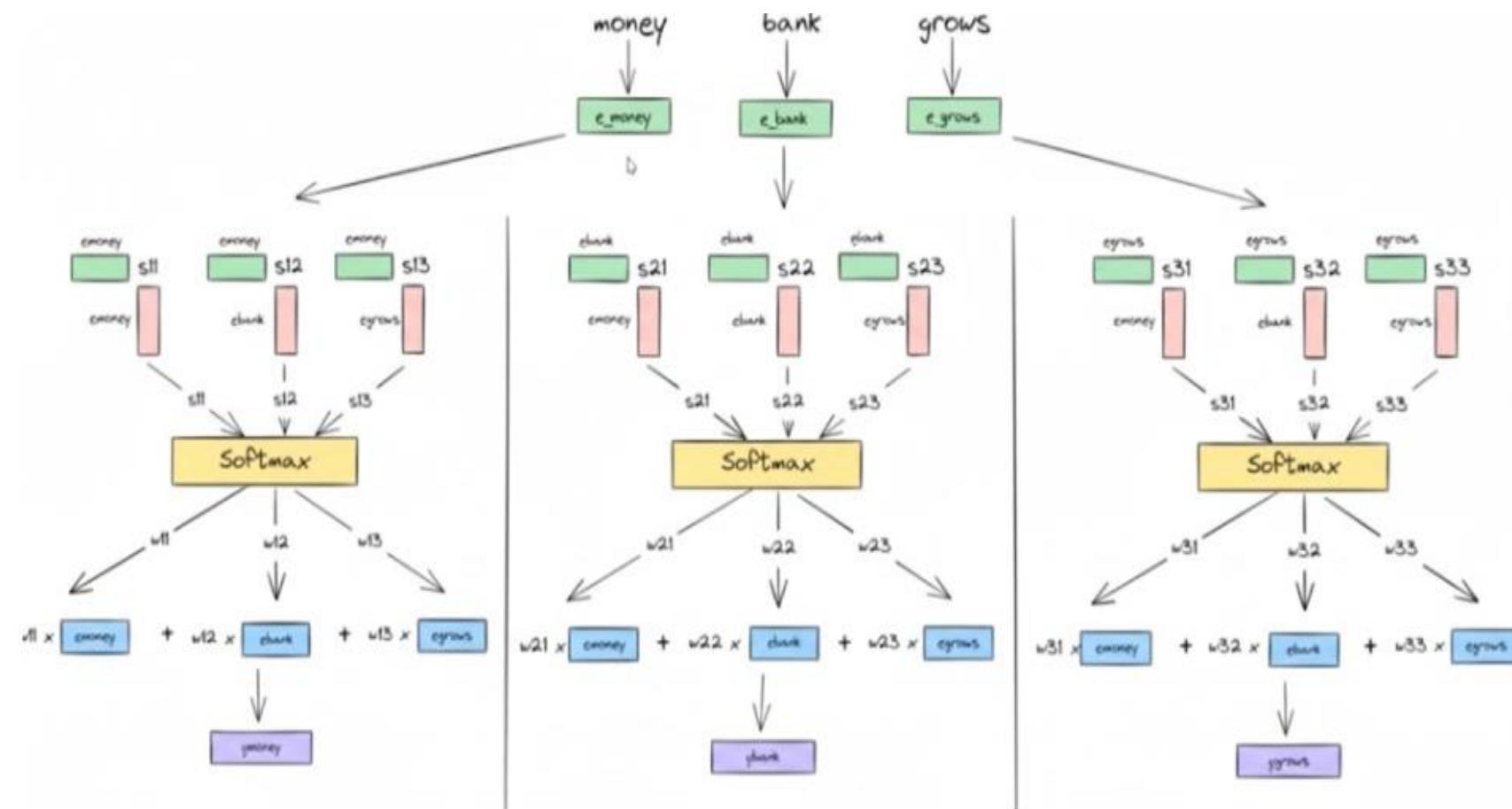
learnable
parameters
general

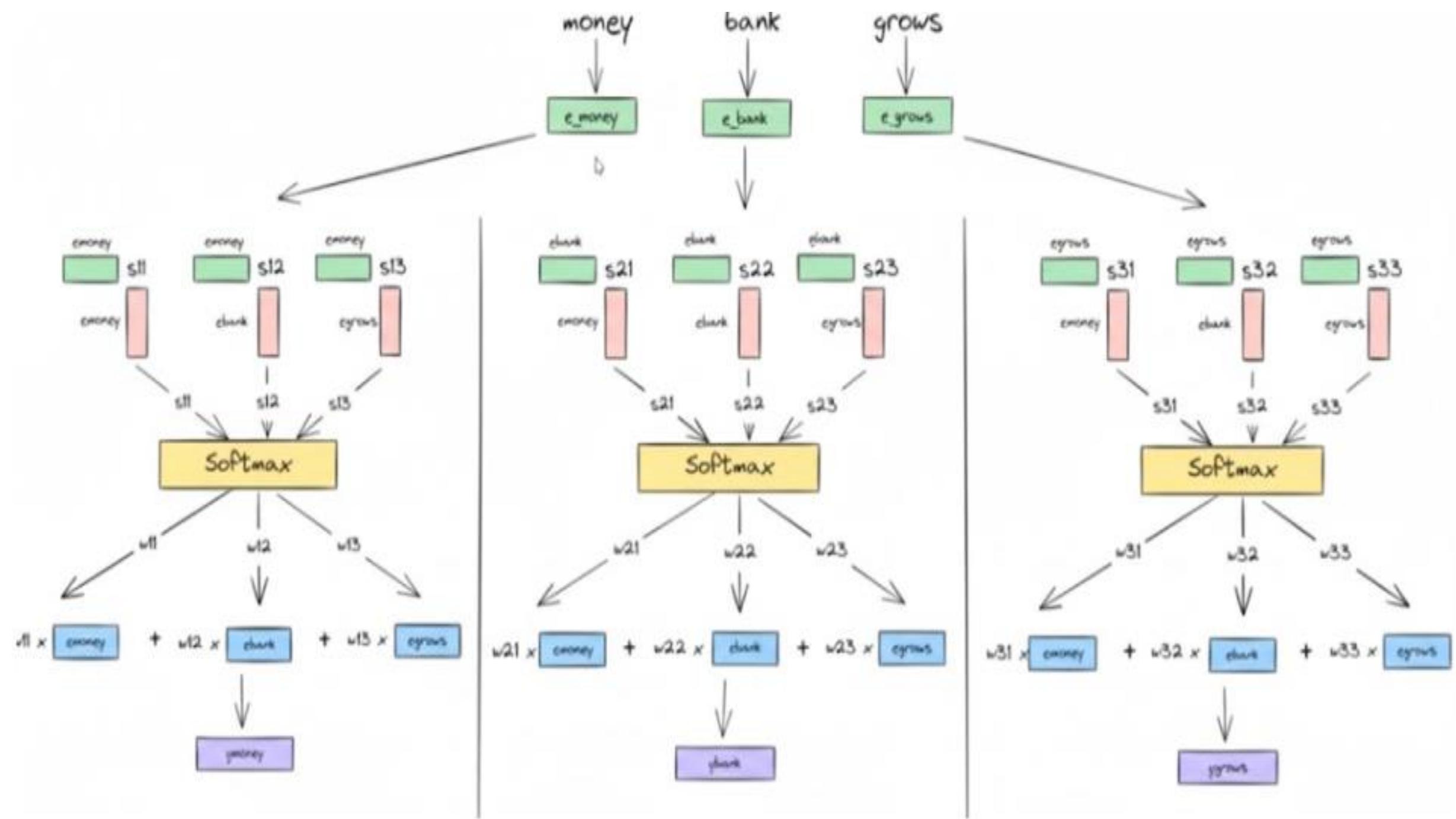
task-specific

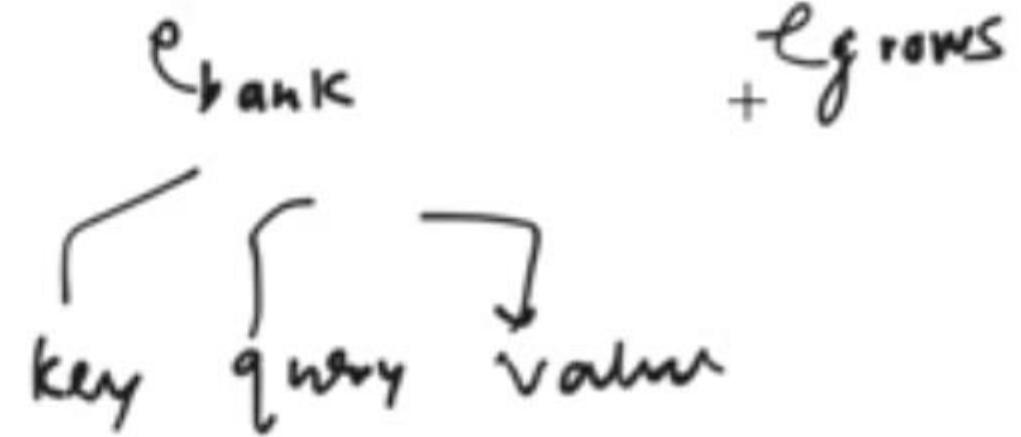
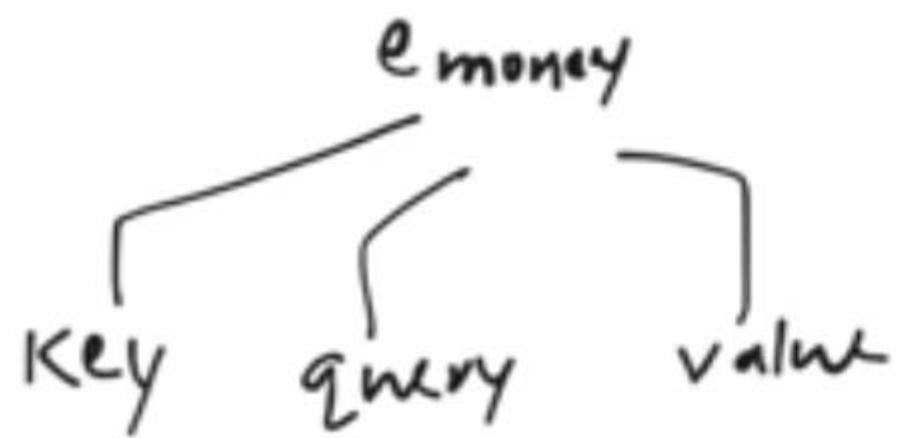
To make it Task specific , we have to add some learnable parameters

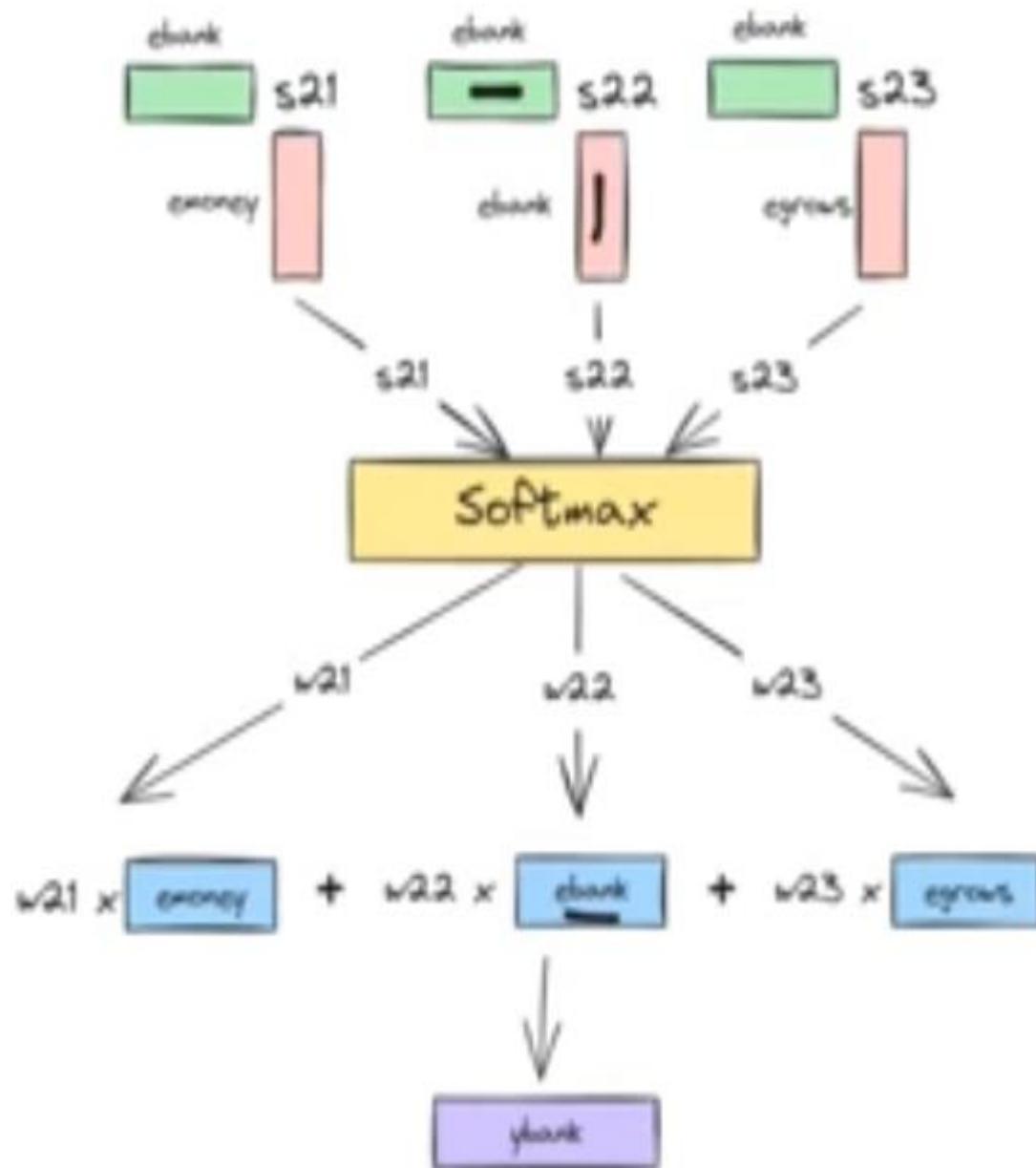
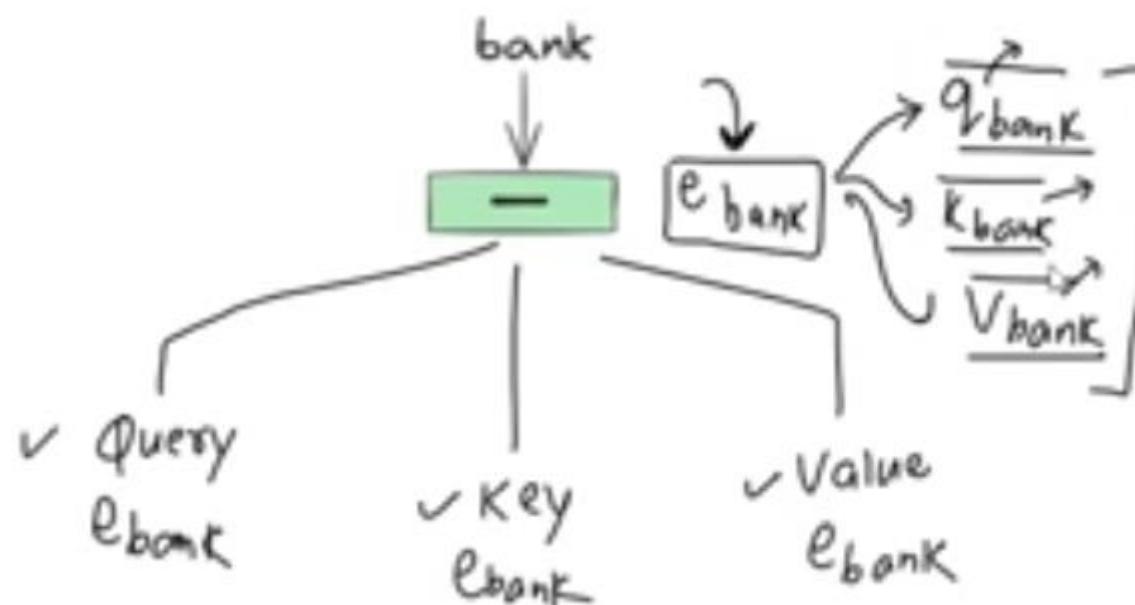


- Where will you add parameter









Separation of Concerns

Person: Author/writer of books → autobiography Ebank

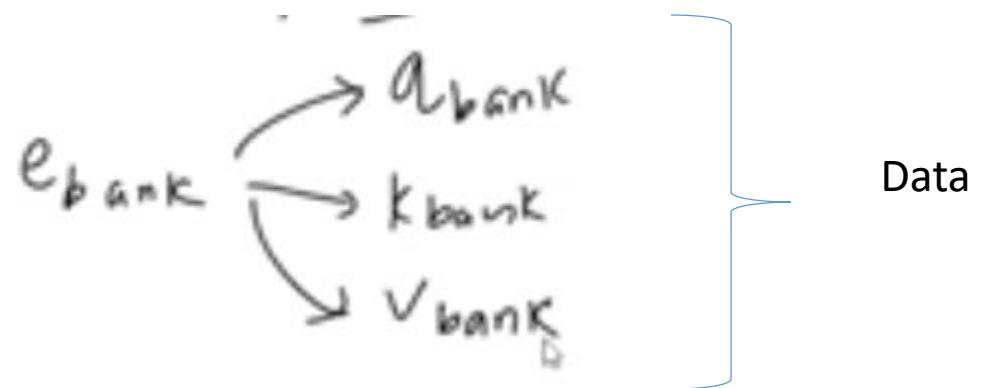
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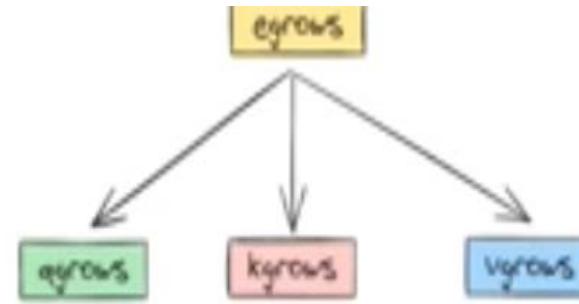
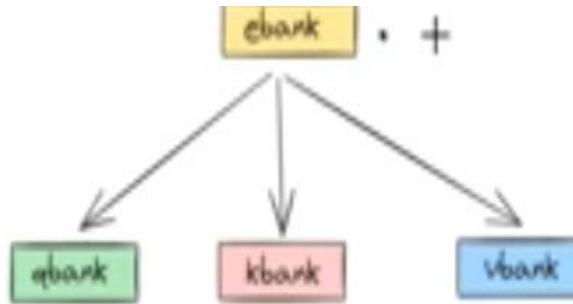
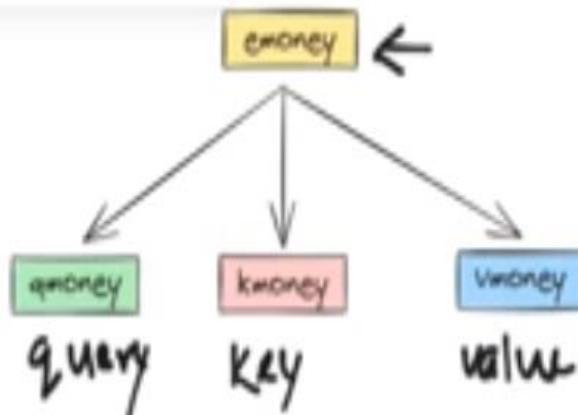
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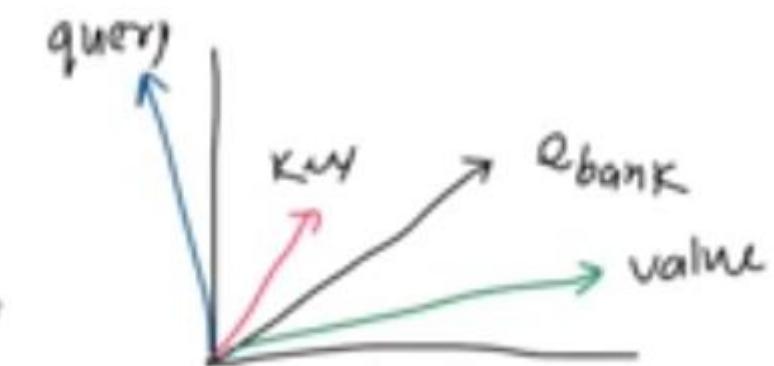
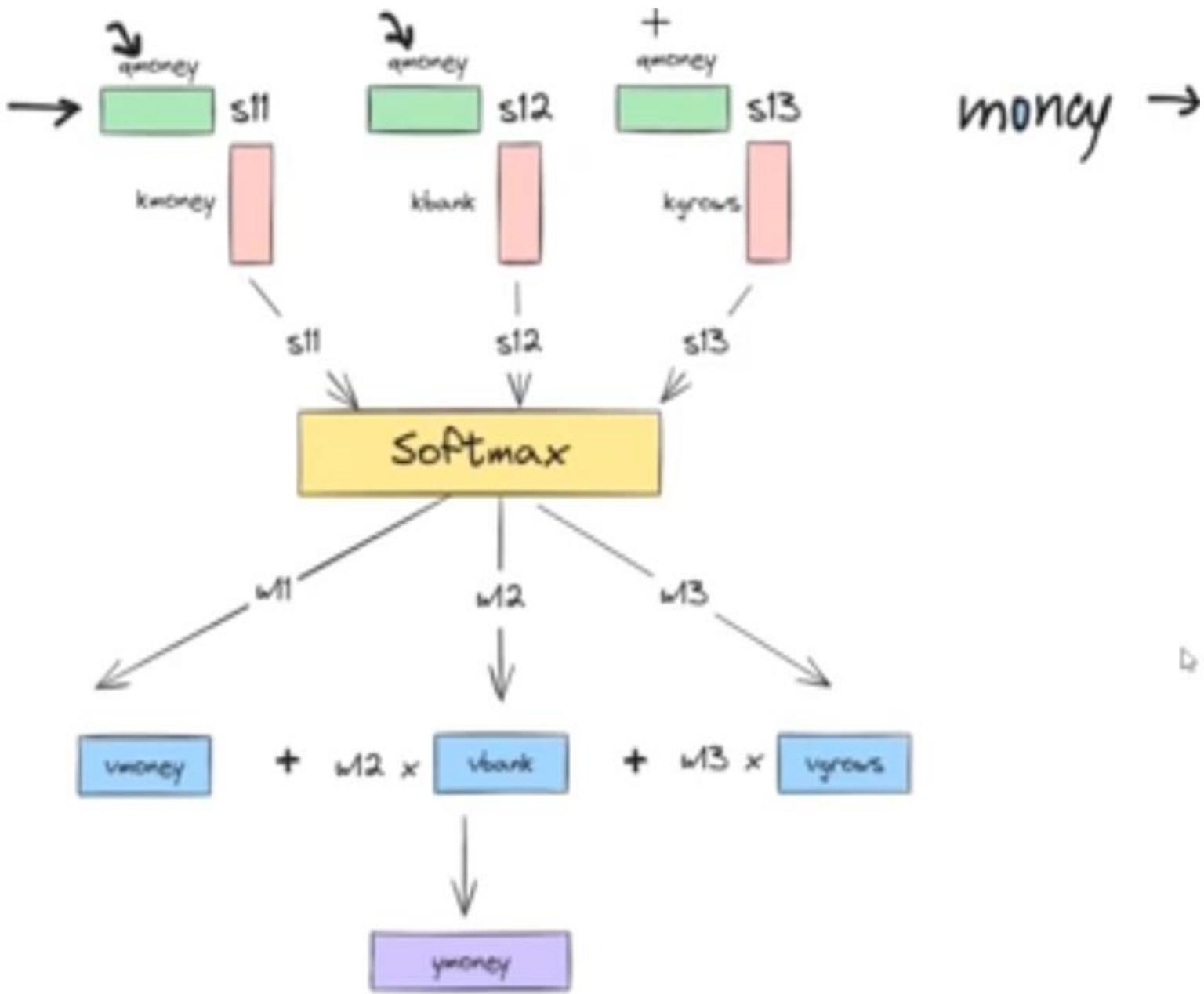
profile
specification/requirements
search(query)
match
meeting/call

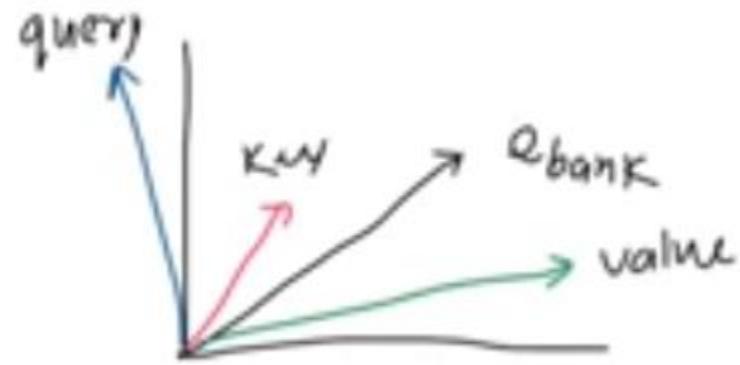
key
Query
value

Autobiography
profile
search
match









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

