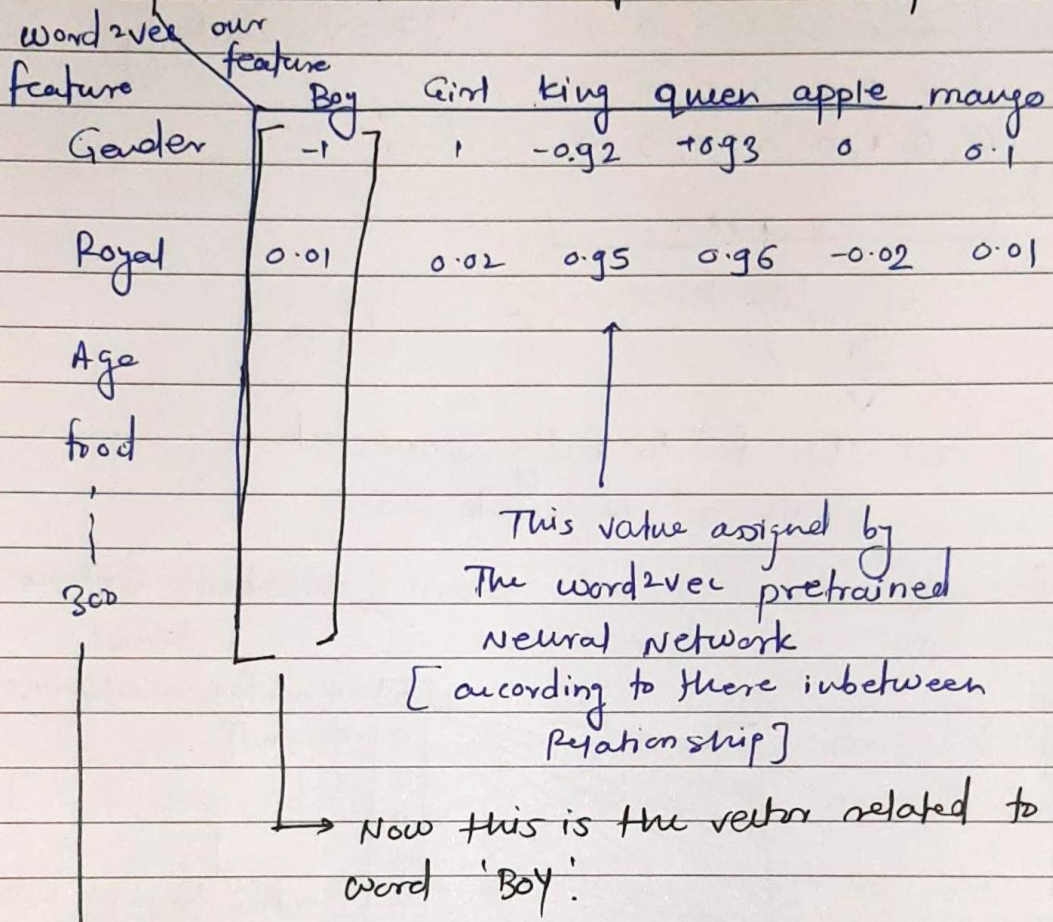


* Word2Vec \rightarrow

- every word has a limited dimension vector.
- Sparsity is reduced.
- Semantic meaning is maintained.
- dense vector is created in word to vec.
- Input for word2vec is simple iterable corpus [.,.,.]



check \rightarrow we don't know the ~~exact~~ exact features over here but we can define the number of the features.

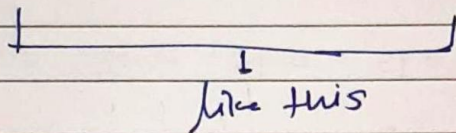
* Continuous Bag of Words:

Ex: [krish channel is Related to Data Science]
suppose window size is = 5
Then

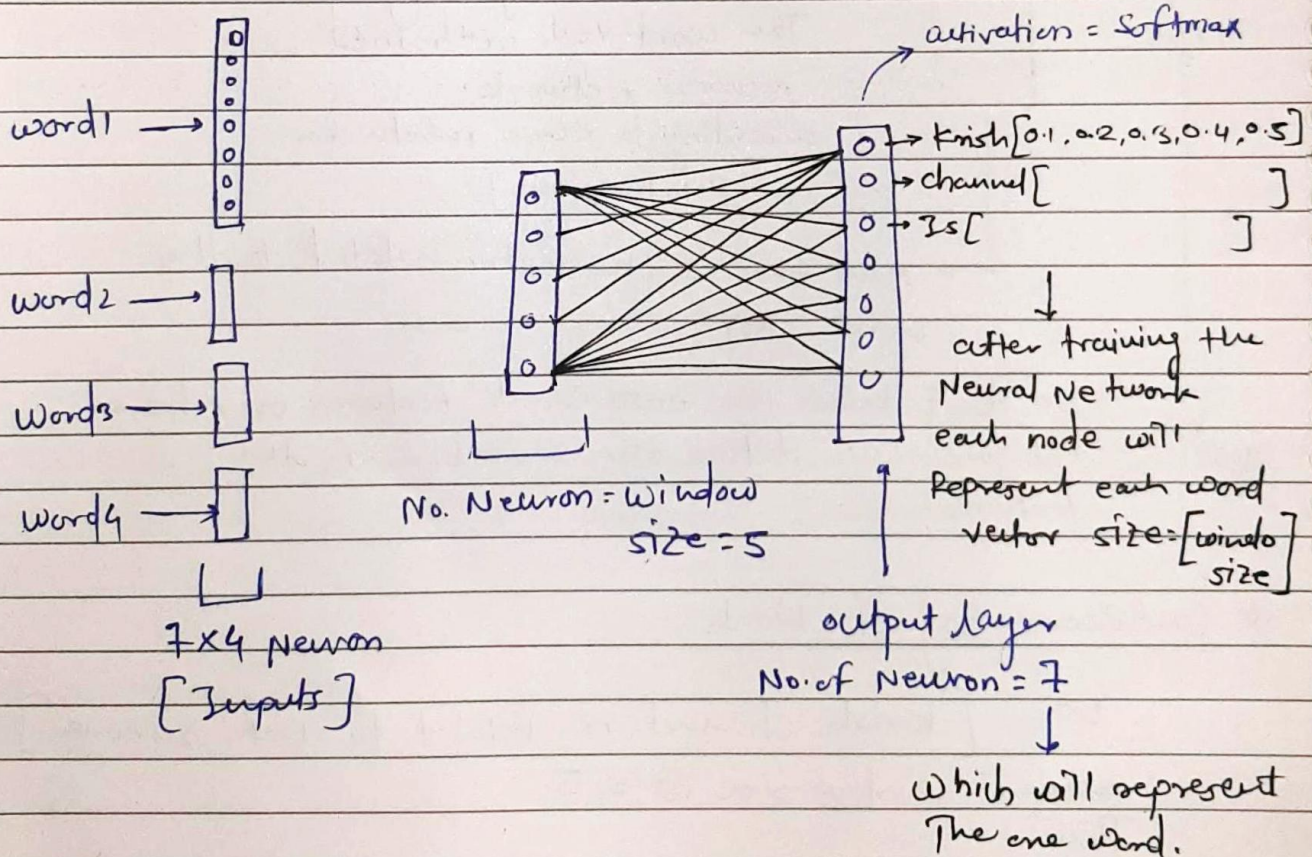
Independent feature	o/p
→ krish, channel, related, to	IS
→ channel, is, to, data	Related
→ Is, related, data, science	To

In Bow How our feature look like

krish → 1, 0, 0, 0, 0, 0, 0, 0 ∴ 7 = Vocabulary
channel 0, 1, 0, 0, 0, 0, 0, 0
is 0, 0, 1, 0, 0, 0, 0, 0



all this fed to fully Connected Neural Network.

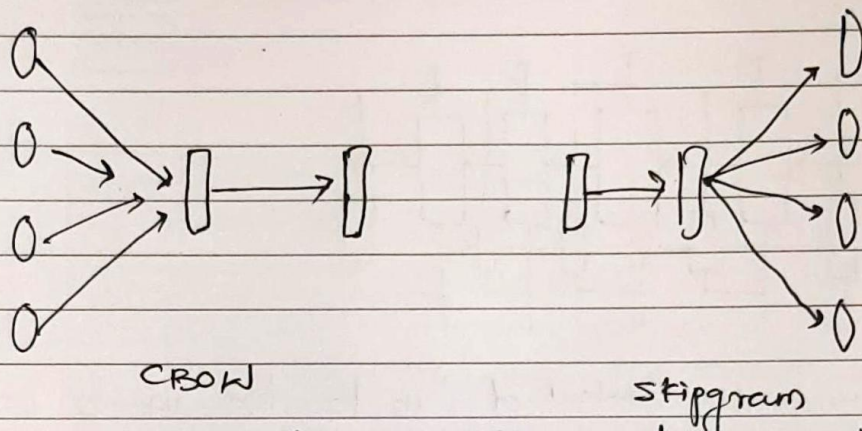


Skipgram \Rightarrow

- This is same process as CBOW but in reverse where

Input	output
Is	crish, channel, related, to.
Related	channel, is, to, data.
To	is related data Science.

Now



- In the CBOW model, the distributed representation of the Context (or Surrounding words) are Combined to predict the word in the middle.
- while in skip-gram model, the distributed representation of the input word is used to predict the Context.

* When to use skip-gram and when to use CBOW?

- ① CBOW \Rightarrow ① if you have larger dataset and you're primarily interested in capturing the meaning of frequent words effectively.
- ② when you have limited computational resources or when training time is concern.
- ② Skip-gram \Rightarrow ① if you have a smaller dataset or you want to capture more information from less frequent words.
- ② when you want ^{capture} their context accurately, regardless of the training time required.