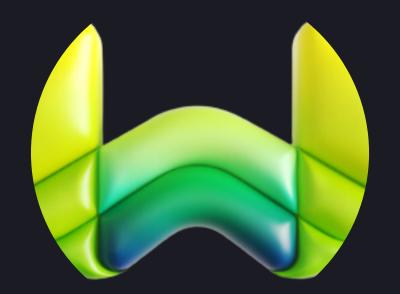
### Weaviate's Guide to

### VECTOR QUANTIZATION

# Every Data Scientist Should Know





Project scaling.

Data accumulating.

Momory usage growing.

Retrieval slowing.

Cost Skyrocketing.

Sounds familiar?



We are in the same boat.

But, there is a solution

### VECTOR QUANTIZATION



### VECTOR QUANTIZATION

- Cuts down memory needs
- Reduces latency
- Slashes cost



# 

Offers solution

in 2 ways



1

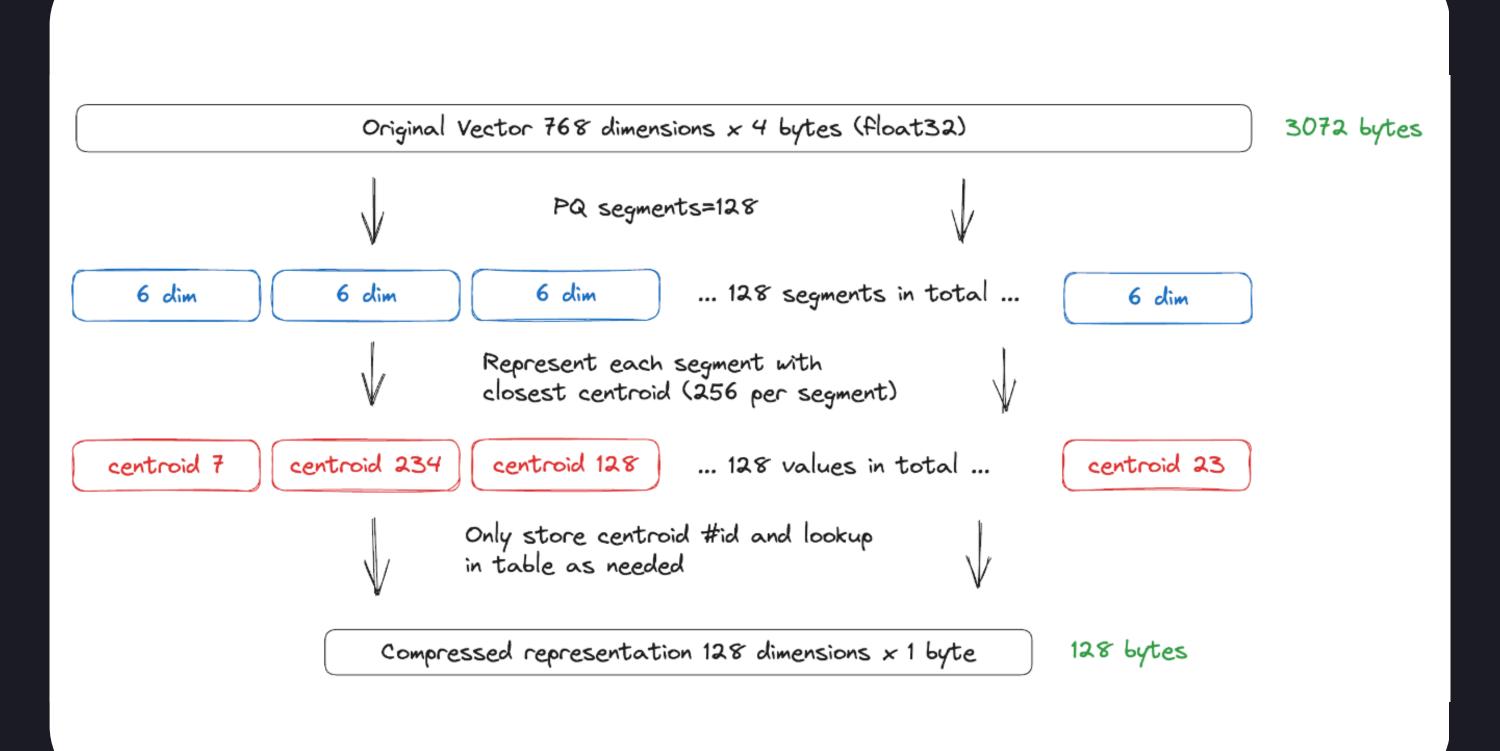
### Product Quantization

(PQ)



#### What PQ does

Compresses your vector embeddings by breaking them down into smaller, manageable segments.





#### PQ Benefits

Reduces memory usage by almost 24 times while maintaining a balance between performance and recall.

#### **Best for**

Those who use hnsw indexes and need a fine balance between speed and accuracy.



2

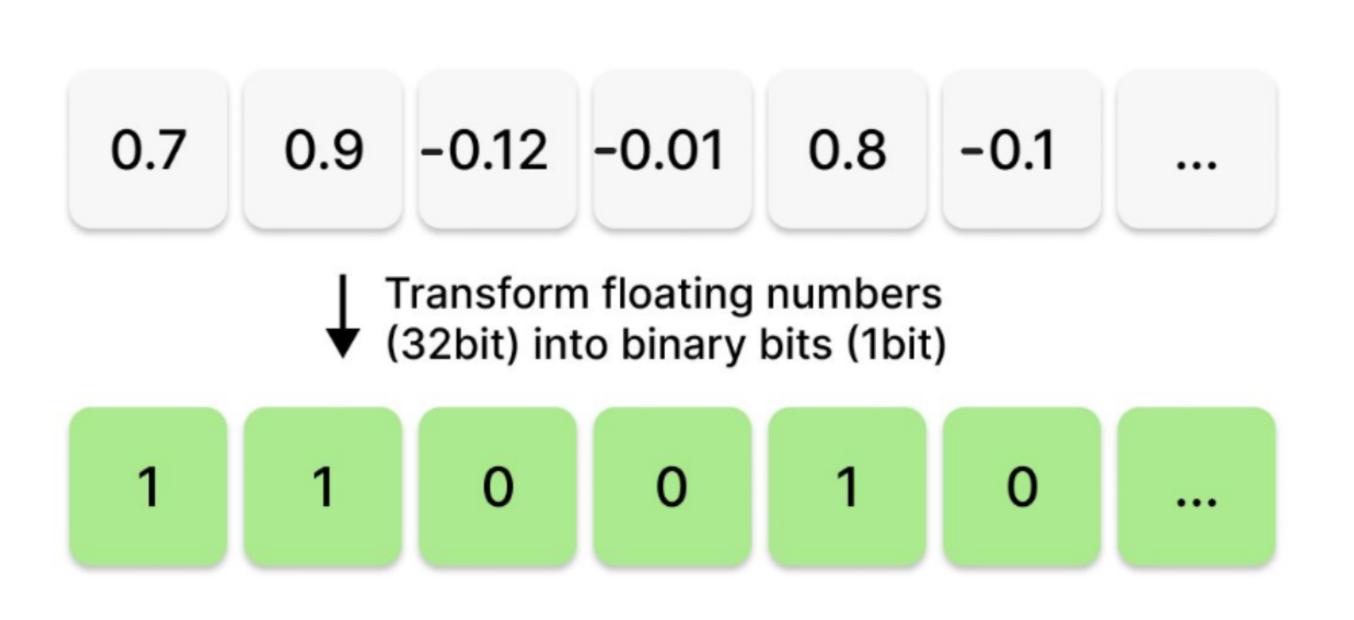
# Binary Quantization

(BQ)



#### What BQ does

Converts each vector into a binary format, drastically reducing the size from bytes to bits.





#### **BQ** Benefits

Achieves a 32x reduction in storage requirements and speeds up search processes.

#### **Best for**

Projects where speed is critical, and slight compromises on accuracy are acceptable.



### Trade-offs

 PQ might slightly reduce recall but saves more memory.

• BQ offers incredible speed at the cost of some accuracy.



# Bonus

Check links in the comment





# Want more content like this?

Follow Qendel AI for daily tips on

Prompting

LLMs

RAG

Agents