



Understanding RAG chunking through simple language, easy math, and playful AI demos!

20 Essential RAG Chunking Methods Every AI Engineer Must Know



Sanjay N Kumar

Data scientist | AI ML Engineer | Statistician | Analytics Consultant

What is Chunking?



Chunking means splitting large text into smaller parts (chunks) so AI can understand better.

📖 Like dividing a long story into chapters!
Without chunking → AI forgets the beginning.
With chunking → AI remembers what matters.

Why Bad Chunking Fails



Bad chunking = AI confusion 

- Breaks sentences halfway
- Loses topic meaning
- Gives wrong answers

Example:

If you cut “*The cat sat on the mat.*” → into “*The cat s*” and “*at on the mat*” – meaning is lost!

Why Good Chunking Helps



Good chunking = AI clarity 

- Keeps sentences complete
- Maintains flow
- Gives correct answers

Example:

Reading one full paragraph at a time makes sense — not random half lines!

The 20 Chunking Techniques



We'll learn 20 common methods:

Fixed-size

Sentence

Paragraph

Sliding window

Semantic

Recursive

Structure-aware

Code block

Query-aware dynamic

Hybrid

The 20 Chunking Techniques



 Context-enriched

 Token-based

 Agentic

 Page-based

 Table-aware

 Image-associated

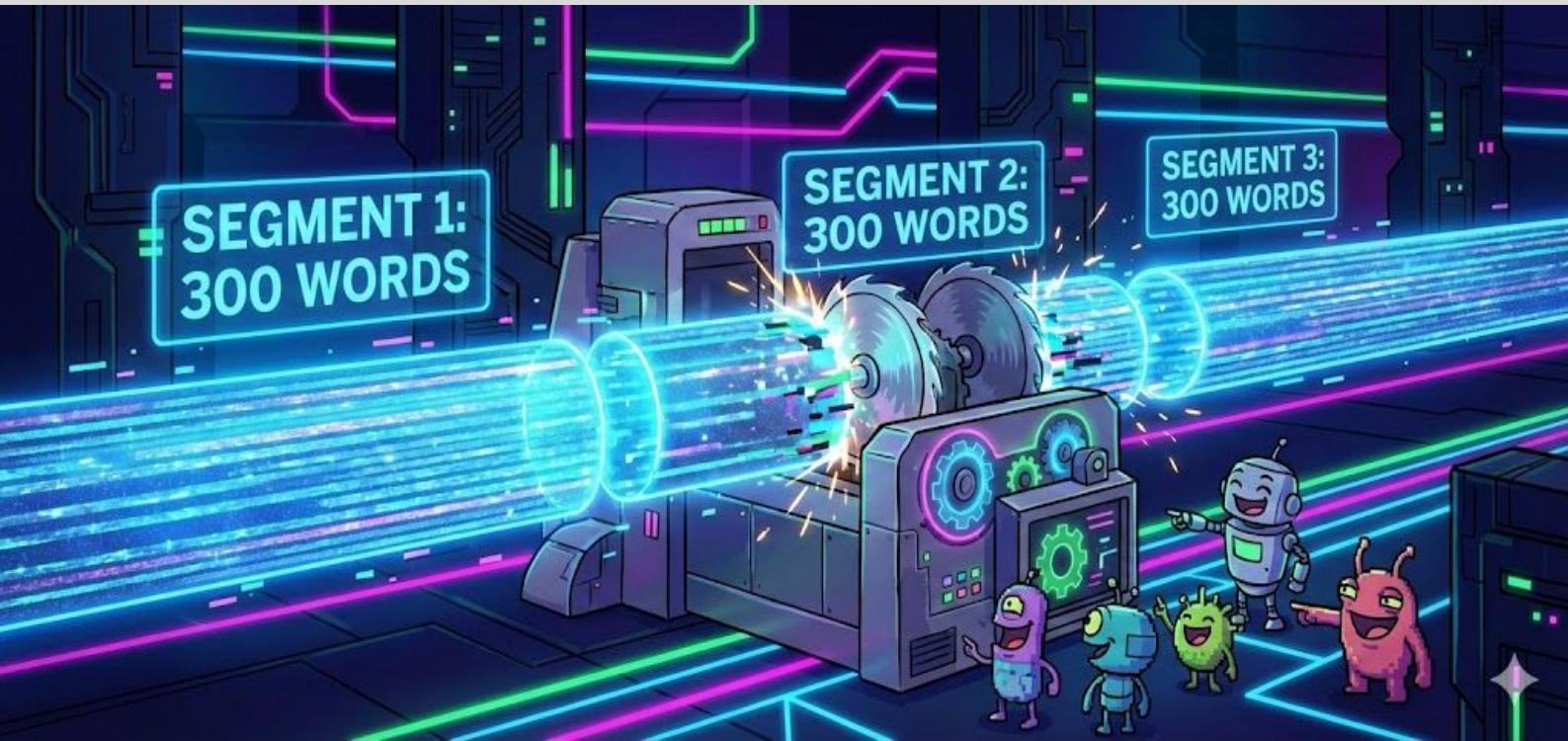
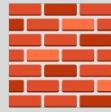
 Event-based

 Citation-aware

 Hierarchical

 Audio/Text alignment

1. Fixed-Size Chunking



Text is split into **equal-sized parts** (like every 300 words).



Simple, fast



May break meaning mid-sentence

Example:

A 900-word story → split into 3 equal 300-word parts.

2. Sentence-Based Chunking 🖊



Each chunk = one full sentence.

✓ Keeps meaning complete

✗ May lose wider context between sentences

Example:

“Dogs bark.” → Chunk 1

“Cats meow.” → Chunk 2

3. Paragraph-Based Chunking



Splits text by paragraph breaks.

✓ Natural separation of ideas

✗ Long paragraphs may exceed limits

Example:

Paragraph 1: “Plants make food.” 

Paragraph 2: “Animals eat plants.” 

4. Sliding Window Chunking



Chunks overlap slightly to keep flow.

✓ Keeps context

✗ Repeats some text

Example:

Chunk 1: Lines 1–5

Chunk 2: Lines 4–8 (overlaps 4–5)

5. Semantic Chunking 🧠



AI splits text by **meaning or topic**.

✓ Very accurate

✗ Slower

Example:

Text about “Earth 🌎” and “Mars 🌈” →
becomes two chunks by topic.

6. Recursive Chunking



Splits large text step by step:

Document → Section → Paragraph →
Sentence.

✓ Flexible

✗ Needs document structure

Example:

Book → Chapter → Paragraph → Line

7. Structure-Aware Chunking



Uses **document structure** like headings, lists, or tags.

- ✓ Best for websites or reports
- ✗ Needs clean formatting

Example:

HTML `<h1>`, `<h2>`, `<p>` → become separate chunks.

8. Code Block Chunking



Splits text by **functions, classes, or methods** in code.

✓ Keeps logic complete

✗ Hard for mixed text+code docs

Example:

Chunk 1: `def add_numbers()`

Chunk 2: `def subtract_numbers()`

9. Query-Aware Dynamic Chunking



AI changes chunking style **based on your question.**

✓ Most relevant results

✗ Complex to build

Example:

If you ask “loan process,” it reads only loan-related chunks.

10. Hybrid Chunking ⚙



Combines two or more methods.

✓ Best of both worlds

✗ Setup is complex

Example:

Use *semantic* + *sliding window* → meaning + context together.

11. Context-Enriched Chunking



Adds extra info like **title** or **source** to each chunk.

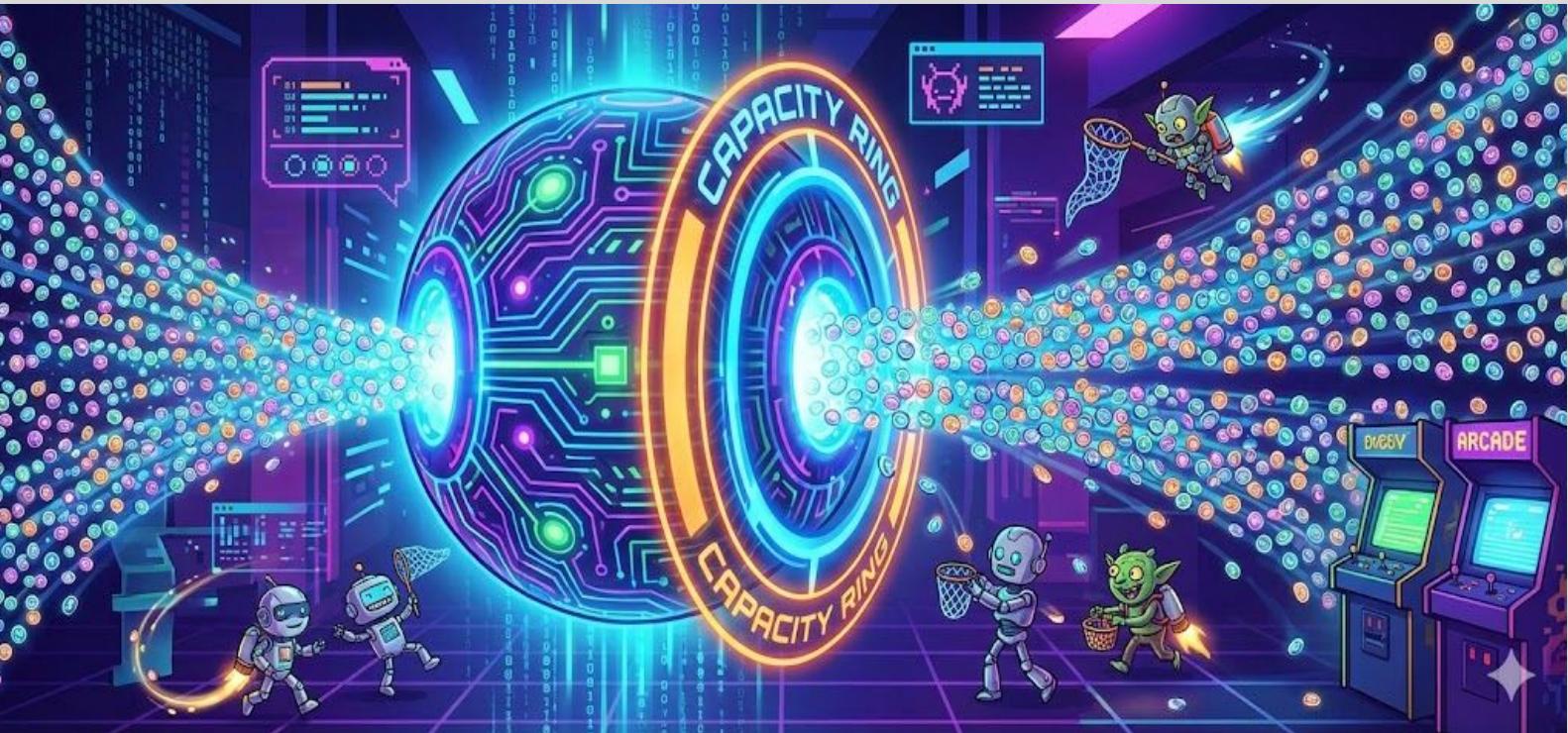
✓ Improves traceability

✗ Uses more tokens

Example:

“Chapter 2 – Photosynthesis 🌱: Plants make food from light.”

12. Token-Based Chunking



AI reads “tokens,” not words.

Splits exactly by token limit (like 512 or 1024).

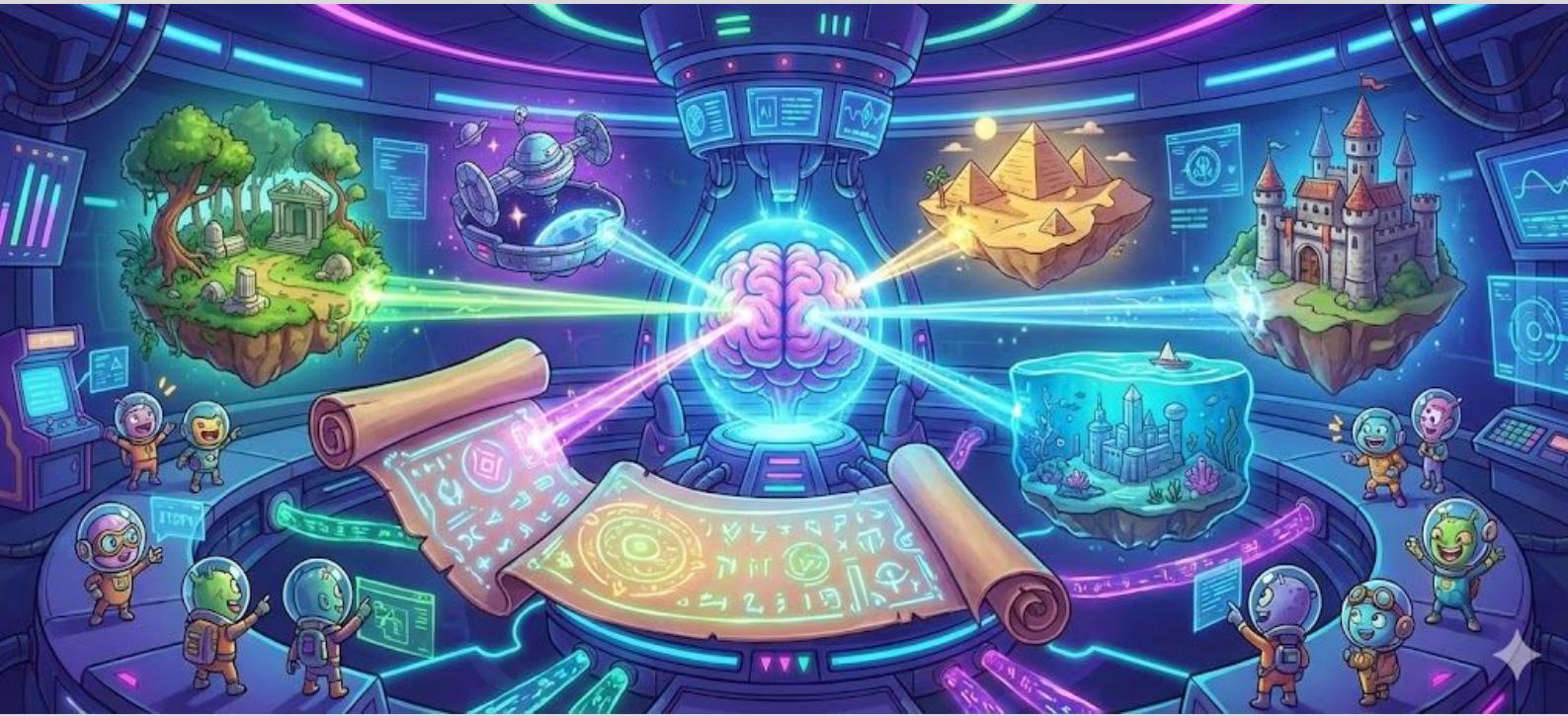
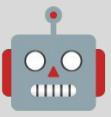
✓ Prevents overflow

✗ May break grammar

Example:

Model fits 1000 tokens → each chunk = 1000-token portion.

13. Agentic Chunking



AI itself decides where to cut based on structure or meaning.

- ✓ Smart and adaptive
- ✗ Computationally heavy

Example:

LLM splits a 20-page report into logical 5-topic groups.

14. Page-Based Chunking



Each **page of a PDF or scan** is one chunk.

- ✓ Keeps layout and numbering
- ✗ Not meaning-based

Example:

Page 1 → Chunk 1

Page 2 → Chunk 2

15. Table-Aware Chunking



Splits tables logically by **rows, columns, or headers**.

✓ Great for data-heavy files

✗ Needs pattern detection

Example:

Chunk 1: Rows 1–5

Chunk 2: Rows 6–10

16. Image-Associated Chunking



Pairs images with nearby text or captions.

✓ Perfect for multimodal RAG

✗ Requires OCR or captioning

Example:

Image of a volcano 🌋 + text “Mount Fuji eruption” → 1 chunk

17. Event-Based Chunking



Splits text by **time or event markers** (like timestamps).

✓ Keeps chronological flow

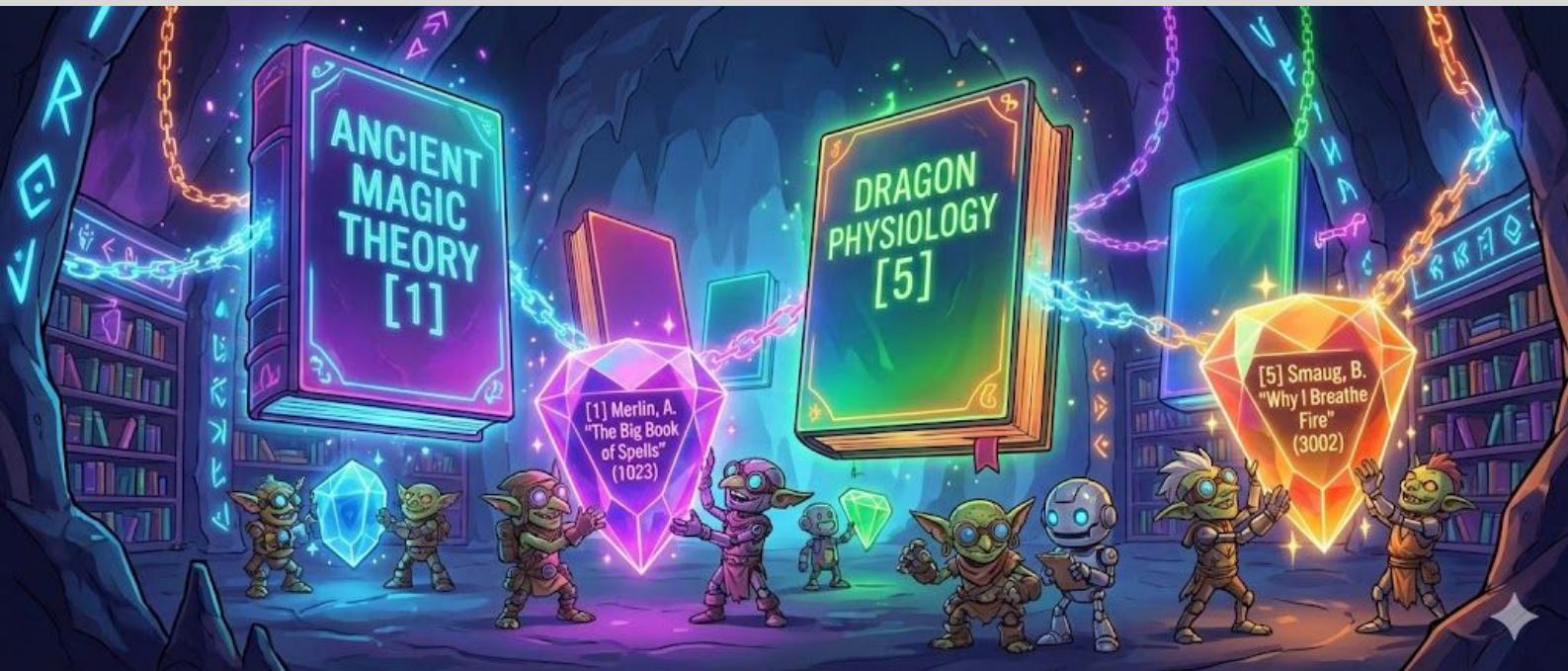
✗ Needs structured logs or transcripts

Example:

[10:00] Meeting start → Chunk 1

[10:30] Discussion → Chunk 2

18. Citation-Aware Chunking



Keeps **references and citations** with the related text.



Useful for legal or research papers

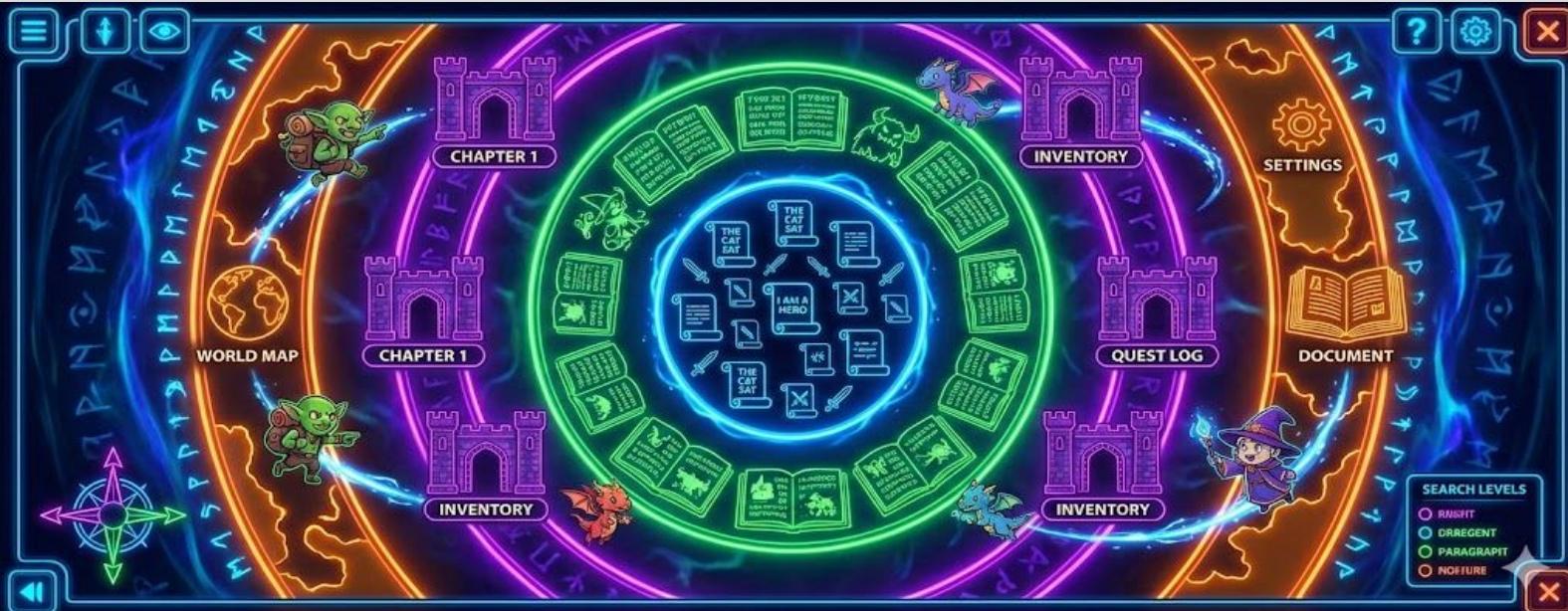


Increases chunk size

Example:

“Quantum theory [Einstein, 1905] explains energy behavior.”

19. Hierarchical Chunking



Creates multi-level chunks:

Sentence → Paragraph → Section → Document.

✓ Supports flexible retrieval

✗ Complex to manage

Example:

You can search at “section level” or “paragraph level.”

20. Audio/Text Alignment Chunking



Used in speech data — aligns chunks with timestamps or speakers.



Keeps voice context



Needs audio timing

Example:

Speaker 1 (00:00–01:00) → Chunk 1

Speaker 2 (01:00–02:00) → Chunk 2

Summary Table



Type	Ideal For	Keeps Meaning	Speed
Fixed	Uniform data	✗	⚡
Sentence	Articles	✓	⚡
Paragraph	Reports	✓	⚡
Sliding	Conversations	✓ ✓	⚡
Semantic	Research	✓ ✓ ✓	⚡
Recursive	Books	✓ ✓	⚡
Structure-Aware	Websites	✓ ✓	⚡
Code Block	Programs	✓	⚡
Query-Aware	Q&A Systems	✓ ✓	⚡
Hybrid	Complex docs	✓ ✓ ✓	⚡
Context-Enriched	Knowledge bases	✓ ✓	⚡
Token-Based	Long texts	✓	⚡
Agentic	Smart docs	✓ ✓ ✓	⚡
Page-Based	PDFs	✓	⚡
Table-Aware	Spreadsheets	✓	⚡
Image-Associated	Visual data	✓	⚡
Event-Based	Logs, transcripts	✓ ✓	⚡
Citation-Aware	Legal, academic	✓ ✓ ✓	⚡
Hierarchical	Layered data	✓ ✓ ✓	⚡
Audio/Text	Meetings	✓ ✓	⚡

Final Takeaway



🧠 Chunking is how you teach your AI *where to pause and think.*

If you chunk wrongly — AI confuses facts.

If you chunk wisely — AI becomes accurate, calm, and smart. 🌟

Fuel Your RAG. Shape the Future



Every chunk you create writes tomorrow's intelligent answers.



Build once. Retrieve forever.

Ready to train smarter? Let's connect and optimize the future!



Sanjay N Kumar

Data scientist | AI ML Engineer | Statistician | Analytics Consultant



<https://www.linkedin.com/in/sanjaytheanalyst360/>



sanjaytheanalyst360@gmail.com