



10

— TYPES OF —

**CHUNKING TECHNIQUES
FOR TEXT PROCESSING
WITH PRACTICAL
IMPLEMENTATION**

```
In [ ]: pip install PyPDF2
```

```
In [ ]: from PyPDF2 import PdfReader

def extract_text_from_pdf(pdf_path):
    reader = PdfReader(pdf_path)
    text = ""
    for page in reader.pages:
        text += page.extract_text()
    return text

pdf_path = "C:\\Users\\nares\\OneDrive\\Desktop\\ml.pdf"
text = extract_text_from_pdf(pdf_path)
print(text)
```

1. Fixed Chunking

Splits text into equal-sized chunks (by word count).

```
In [ ]: def fixed_chunking(text, chunk_size=100):
        words = text.split()
        chunks = [' '.join(words[i:i + chunk_size]) for i in range(0, len(words)
        return chunks

chunks = fixed_chunking(text, chunk_size=10)
for i, chunk in enumerate(chunks):
    print(f"Chunk {i+1}: {chunk}")
```

2. Overlapping Chunking

Splits text into chunks with an overlap for smoother transitions.

```
In [ ]: def overlapping_chunking(text, chunk_size=100, overlap=50):
        words = text.split()
        chunks = [' '.join(words[i:i + chunk_size]) for i in range(0, len(words)
        return chunks

chunks = overlapping_chunking(text, chunk_size=20, overlap=5)
for i, chunk in enumerate(chunks):
    print(f"Chunk {i+1}: {chunk}")
```

3. Semantic Chunking

Uses spaCy to split text into meaningful sentences.

```
In [ ]: pip install spacy
```

```
In [ ]: import spacy.cli
        spacy.cli.download("en_core_web_sm")

        def semantic_chunking(text):
            nlp = spacy.load("en_core_web_sm")
            doc = nlp(text)
            chunks = [sent.text for sent in doc.sents]
            return chunks

        chunks = semantic_chunking(text)
        for i, chunk in enumerate(chunks):
            print(f"Chunk {i+1}: {chunk}")
```

4. Recursive Character Chunking

Splits text recursively based on character count, prioritizing word boundaries.

```
In [ ]: def recursive_chunk(text, max_size):
        if len(text) <= max_size:
            return [text]
        split_point = text.rfind(" ", 0, max_size)
        if split_point == -1: # No space found, force split
            split_point = max_size
        chunk = text[:split_point]
        remaining_text = text[split_point:].strip() # Remove leading spaces
        return [chunk] + recursive_chunk(remaining_text, max_size)

        chunks = recursive_chunk(text, 100)
        for i, chunk in enumerate(chunks):
            print(f"Chunk {i+1}: {chunk}")
```

5. Agentic Chunking

Uses an AI agent (via Groq API) to split text meaningfully for a given task.

```
In [ ]: pip install groq
```

```
In [ ]: from groq import Groq

        def agentic_chunking(text, task="summarize"):
            client = Groq(api_key="your_api_key")
            prompt = f"Split the following text into meaningful chunks for the task"
            response = client.chat.completions.create(
                model="llama3-70b-8192",
                messages=[{"role": "user", "content": prompt}]
            )
            chunks = response.choices[0].message.content.split('\n')
            return chunks

        chunks = agentic_chunking(text, task="summarize")
```

```
for i, chunk in enumerate(chunks):
    print(f"Chunk {i+1}: {chunk}")
```

6. Advanced Semantic Chunking

Uses SentenceTransformer and KMeans clustering to group semantically similar sentences.

```
In [ ]: !pip install sentence-transformers scikit-learn numpy
```

```
In [ ]: from sentence_transformers import SentenceTransformer
from sklearn.cluster import KMeans
import numpy as np

def advanced_semantic_chunking(text, num_chunks=15):
    model = SentenceTransformer('all-MiniLM-L6-v2')
    sentences = text.split('.')
    embeddings = model.encode(sentences)
    kmeans = KMeans(n_clusters=num_chunks)
    kmeans.fit(embeddings)
    clusters = kmeans.labels_
    chunks = [[] for _ in range(num_chunks)]
    for i, cluster in enumerate(clusters):
        chunks[cluster].append(sentences[i])
    return [' '.join(chunk) for chunk in chunks]

chunks = advanced_semantic_chunking(text, num_chunks=5)
for i, chunk in enumerate(chunks):
    print(f"Chunk {i+1}: {chunk}")
```

7. Context Enriched Chunking

Combines surrounding sentences to add context to each chunk.

```
In [ ]: def context_enriched_chunking(text, window_size=2):
    sentences = text.split('.')
    chunks = []
    for i in range(len(sentences)):
        start = max(0, i - window_size)
        end = min(len(sentences), i + window_size + 1)
        chunk = ' '.join(sentences[start:end])
        chunks.append(chunk)
    return chunks

chunks = context_enriched_chunking(text, window_size=1)
for i, chunk in enumerate(chunks):
    print(f"Chunk {i+1}: {chunk}")
```

8. Paragraph Chunking

Splits text based on paragraphs (using double line breaks).

```
In [ ]: def paragraph_chunking(text):
        paragraphs = text.split('\n\n')
        return paragraphs

chunks = paragraph_chunking(text)
for i, chunk in enumerate(chunks):
    print(f"Chunk {i+1}: {chunk}")
```

9. Recursive Sentence Chunking

Recursively splits text into chunks based on a set number of sentences.

```
In [ ]: def recursive_sentence_chunking(text, max_sentences=3):
        sentences = text.split('.')
        if len(sentences) <= max_sentences:
            return [' '.join(sentences)]
        chunk = ' '.join(sentences[:max_sentences])
        remaining = ' '.join(sentences[max_sentences:])
        return [chunk] + recursive_sentence_chunking(remaining, max_sentences)

chunks = recursive_sentence_chunking(text, max_sentences=3)
for i, chunk in enumerate(chunks):
    print(f"Chunk {i+1}: {chunk}")
```

10. Token Based Chunking

Splits text into chunks based on a specific token count.

```
In [ ]: def token_based_chunking(text, token_limit=50):
        tokens = text.split() # Basic tokenization by whitespace
        chunks = [' '.join(tokens[i:i+token_limit]) for i in range(0, len(tokens), token_limit)]
        return chunks

chunks = token_based_chunking(text, token_limit=50)
for i, chunk in enumerate(chunks):
    print(f"Chunk {i+1}: {chunk}")
```