1.0 Core Strategy: Rule-Based Heuristic Approach

Our primary strategy is to build a high-performance, rule-based system. This approach analyzes the stylistic and positional properties of text within the PDF to identify headings. It avoids the overhead and size of machine learning models, ensuring we meet the hackathon's strict constraints on execution time (≤ 10 s) and size (≤ 200 MB).

Key Library: PyMuPDF (fitz) will be the core of our solution. It is a powerful Python library known for its speed and ability to extract detailed text metadata, such as font size, font name, bold flags, and coordinates.

2.0 Step-by-Step Implementation Plan

2.1 Environment Setup

The solution will be containerized using Docker. The project structure will include a Dockerfile, a requirements.txt file, and the main Python script.

File: requirements.txt PyMuPDF==1.24.1

File: Dockerfile

Use a lightweight, amd64-compatible Python base image FROM --platform=linux/amd64 python:3.10-slim

Set the working directory WORKDIR /app

Copy dependency list and install them COPY requirements.txt . RUN pip install --no-cache-dir -r requirements.txt

Copy the solution code into the container COPY . .

The command to run your solution. This script will handle the I/O logic. CMD ["python", "main.py"]

2.2 Main Script (main.py)

This script will manage the file I/O operations, processing all PDFs from /app/input and writing JSON results to /app/output.

File: main.py import os import json import fitz # PyMuPDF

Define input and output directories INPUT_DIR = "/app/input"
OUTPUT_DIR = "/app/output"

```
# The core logic will be in this function
def extract_outline_from_pdf(pdf_path):
  doc = fitz.open(pdf path)
  title = doc.metadata.get('title', 'Untitled')
  blocks = []
  # 1. Collect all text blocks with their properties
  for page num, page in enumerate(doc):
     page_blocks = page.get_text("dict")["blocks"]
     for block in page blocks:
       if "lines" in block:
          for line in block["lines"]:
             for span in line["spans"]:
               blocks.append({
                  "text": span["text"].strip(),
                  "size": span["size"],
                  "font": span["font"],
                  "page": page num + 1
               })
  if not blocks:
     return {"title": title, "outline": []}
  # Find the most common font size (likely body text)
  sizes = [b['size'] for b in blocks]
  most common size = max(set(sizes), key=sizes.count)
  # 2. Filter for potential headings
  heading_candidates = [b for b in blocks if b['size'] > most_common_size and b['text']]
  # 3. Identify and rank unique heading styles by size
  unique styles = sorted(list(set(b['size'] for b in heading candidates)), reverse=True)
  style map = {}
  if len(unique_styles) > 0: style_map[unique_styles[0]] = "H1"
  if len(unique styles) > 1: style map[unique styles[1]] = "H2"
  if len(unique_styles) > 2: style_map[unique_styles[2]] = "H3"
  # 4. Build the final outline
  outline = ∏
  for block in heading candidates:
     if block['size'] in style_map:
       outline.append({
          "level": style_map[block['size']],
```

```
"text": block['text'],
          "page": block['page']
       })
  # Use largest font on first page for title if metadata is poor
  if not title or len(title) < 4:
     first page blocks = [b for b in blocks if b['page'] == 1]
     if first page blocks:
       title = max(first_page_blocks, key=lambda x: x['size'])['text']
  return {"title": title, "outline": outline}
if name == " main ":
  if not os.path.exists(OUTPUT DIR):
     os.makedirs(OUTPUT_DIR)
  for filename in os.listdir(INPUT_DIR):
     if filename.lower().endswith(".pdf"):
       pdf_path = os.path.join(INPUT_DIR, filename)
       outline data = extract outline from pdf(pdf path)
       output filename = os.path.splitext(filename)[0] + ".json"
       output path = os.path.join(OUTPUT DIR, output filename)
       with open(output_path, 'w', encoding='utf-8') as f:
          json.dump(outline data, f, indent=2, ensure ascii=False)
```

3.0 Meeting Constraints and Scoring

- * Execution Time (≤ 10s): The PyMuPDF library is highly optimized and will easily process a 50-page PDF in under a second.
- * Model Size (≤ 200MB): This approach uses no machine learning models, so the constraint is met by default.
- * Network: The solution is entirely self-contained and requires no internet access.
- * Accuracy: Accuracy can be further improved by adding heuristics for font weight (e.g., "Bold" in the font name), text numbering (1.1, A.), and line length.
- * Bonus (Multilingual): PyMuPDF correctly extracts Unicode text. For languages like Japanese, where bold flags are uncommon, the system's reliance on font size as the primary differentiator is a significant advantage.