# Parsing PDFs on a laptop or onpremise

# About this of session/presentation

- part 1 insists on theory and favors knowledge sharing
- part 2 will happen later and will contain more live coding/practical examples
- all presented tools are very well documented you could get started after part 1
- why is part 1 so long: the hardest part is choosing the right tool
   for the right type of PDF
- it's dense: feel free to reach out to me later
- slides are a bit full, not ideal for presenting, but easier to use and share as standalone doc later on

# Part 1: theory

# Why should I even care about local PDF parsing (in the LLM era)?

- LLMs often offer advanced doc parsing capabilities
- however, there are a few good reasons to use alternative approaches when possible

# **Confidentiality**

- in many Data Journalism applications, it's not possible to send data to the cloud for confidentiality reasons
- it's often the case at ICIJ
- local (or on-premise) alternatives to cloud APIs are still very relevant

# Cost control 😝 and ecological impact 🥵

- sometimes, very basic Python libs can perfectly handle PDFs
- given the financial and environmental cost of LLMs  $\P$   $\stackrel{\longleftarrow}{\blacksquare}$ , there's no reason to use them when we can avoid it
- it's all the more important in large-scale applications when dealing with thousands/millions of pages of docs

# Even LLM applications can require text inputs

- some LLMs applications require turning docs into text format first
- it's typically the case when performing **Retrieval-Augmented-Generation (aka <u>RAG</u>)**
- in RAG, docs are first converted to text (ideally Markdown) then indexed and embedded for vector (or hybrid) search
- for RAG: better PDF parsing → better doc embedding → better doc retrieval → better LLM answer

# Avoiding hallucinations and aiming for reproducibility

- LLMs can get creative, they should be avoided when fidelity is key
- even when using a seed + temperature=0 LLMs outputs are often non-deterministic (due to batch processing and floating point operations)
- some level of reproducibility is important when you aim at improving something in a **stable** manner over time

# Wait a bit, what's a PDF file 👺?

# PDFs are meant to be visualized, not parsed

#### PDFs often contain:

- a description of elements to display in the PDF viewer. Think of it as a set of instructions to write/draw the doc content on a whiteboard. It describes what to write/draw and where to do it
- embedded files (optional): hidden inside the PDF file itself, typically images
- metadata: author, name of the software which created the PDF, some security stuff, anything...
- scripts (optional): PDF can contain all sorts of malware 1

# PDF parsing challenges

- PDFs are essentially descriptive
- there are often many ways to write/draw a document on a whiteboard to get to the exact same result: no canonical form
- some PDFs files can be insanely complex
- no simple or canonical way to parse PDFs

## To go further

- Wikipedia PDF article
- <u>Understanding the PDF file format</u>
- PDF file format <a href="RFC">RFC</a> (PDF file format specification)

## The different "kinds" of PDFs

When it comes to parsing PDFs, we can roughly categorize docs in two different categories:

- digital documents (computer generated PDFs)
- scans and images

# Digital documents (computer generated PDFs)

#### Invoice

**A GO RAILS** 

Invoice Number 123

Issue Date March 25, 2024
Due Date March 25, 2024
Status PAID

Example, LLC 123 Fake Street New York City, NY 10012

(555) 867-5309 support@example.com Bill To
Customer
Address
City, State Zipcode
customer@example.org

Item	Unit Cost	Quantity	Amount
Subscription	\$19.00	1	\$19.00
		Subtotal	\$19.00
		Tax Rate	0%
		Amount Due	\$19.00

For questions, contact us anytime at support@example.com.

- homemade term to designate any doc originally created digitally and later saved/exported/converted as PDFs to be shared
- text-oriented docs (often generated from Word, OpenOffice, Google Docs...)
- invoices, forms, contracts... (generated by programs)
- any doc created by a "Save as PDF", "Export as PDF"... functionality

- computer generated PDFs often contain a lot of readable text
- computer generated PDFs are **generally well handled** by PDF parsing libs, performances can greatly vary depending on:
  - the software used to generate the PDF
  - PDF items of interest, while texts in paragraphs are usually easily read, other items are not trivial to parse even in digital docs:
    - tables
    - diagrams
    - embedded images
    - more generally any vectorized/raster content

# Scans and images

#### Xerox® VersaLink® C405DN Multifunction Printer



#### Première page

"VersaLink C405" Nom de périphérique 14248 Total impressions Numéro de série 3990203392 26/04/2021 Date d'activation Version 3018.103 PostScript 3 Version 100.1 PCL6/5e

Version du logiciel Version 68.52.1 Version 1.60.0 Controller ROM Version 3.4.0 IOT ROM ADF ROM Version 42.0.0 Fax ROM Version 104.8.0 Version 91.16.11 Panel ROM Version 11.1.216 **Boot ROM** Version 4.7.0 Version 2.5.1 SSMI Contents Version 5.0.0 Version 1.8.25 26/04/2021 15:11 Logiciel mis à jour 2.00 Go RAM installée

136 polices européennes PS Polices ROM disponibles

83 PCL\*

Disque dur / Disque SSD Non connecté Module recto verso Installé Module de finition Non installé Activé Plan consommables Activée USB 3.0 Activée Impression USB 3.0 Numérisation USB 3.0 Activée

Automatique . "9C:93:4E:D3:F9:F6" Ethernet

802.1X Désactivé "192.168.15.50" TCP/IP IPv4 TCP/IP IPv6 Désactivé(e) pltv.local Nom DNS

https://XC-D3F9F6.pltv.local:443 Page Web

Connectivité - Flux de travail mobiles

Google Cloud Print

- contrary to digital docs, they are primarily images converted as
   PDF files
- strictly speaking, scanned docs are also digital docs and computergenerated
- scans and images often contain no text inside we can't just
   read
- from a PDF format perspective are almost like raw images (JPEG, PNG...) converted to PDF
- contrary images converted to PDF, scans often suffer from quality issues

# Why can I select text even in scanned documents?

- this doc was scanned using from my phone, when I opening it in a PDF viewer text can be selected
- By why isn't it a pure image, it looks like a "digital doc"
- when scanning a doc, some software runs OCR to read the text and adds this extra information to display it on the top of the scan
- my phone uses the latest Al during OCR the doc, so when parsing we could just read the OCR output
- in the general case, it's not reliable, it depends on the scanning software, image quality, scan age (old scan → old OCR)

Scanned docs are usually much harder to parse than digital docs / computer generated PDFs

# I have a set of PDFs I need to process, what tool should I use?

Spoiler alert: no single tool to rule them all

#### PDFs handling workflow is speed/cost yes heuristics critical? (level 1) not at all complex layout ML + digital scans or digital yes is speed/cost yes and/or heuristics critical? documents? image items? (level 2) no scans end-to-end ML (level 3)

# **PDF** parsing libraries

# PDF parsing features

Most libs offer two distinct sets of features:

### parsing features

- after parsing, the PDF structure items and their content become accessible
- you can navigate through the PDF elements using Python code, just like you would scrap a webpage: select/filter/access titles, header, footer, table rows and columns... and read their content
- each lib uses its own schema/ontology/language to describe
   PDFs structure (whereas all scraping libs use HTML)

• rendering: convert the whole PDF or some of its items into another other format (Markdown, CSV, JSON, HTML...)

# Level 1: heuristic-based parsing

- heuristics ≈ good old code, logical rules implemented from priori knowledge
- sometimes relies on image processing techniques (<u>OpenCV</u>) to understand layout
- the lib guess PDF structured from the visual description, it tries to guess its structure from **visual clues** 
  - "if there's a large space vertically between text characters, there might a new paragraph"
  - "when there are a lot of parallel vertical lines which are perpendicular to parallel horizontal line, there might be a table"

#### Pros:

- works well for most computer generated PDFs
- fast

#### Cons:

- since there is no canonical way to write/draw on the "whiteboard", heuristics can easily fail
- often not well suited for scans, images (lines are not straight nor parallel anymore)
- often not well suite for PDFs with diagrams/charts, and more generally vectorized information
- often fail on PDFs with complex structure (columns, complex tables...)

## <u>pdfplumber</u>

- low-level lib to access structure and content items (webscraper-like API)
- used to access PDF items not to extract/read their content (no OCR, no advanced layout detection)
- useful when low-level access to PDF structure is required

## **PyMuPDF**

• similar to pdfplumber but faster and hence not implemented in Python (harder to debug)

### **Parsr**

- PDF parser from AXA
- aggregates multiple heuristics from other PDF parsing libs
- can be used to match/detect some elements and then export results as CSV/Markdown
- can run as a local Docker image and called from HTTP

## **Camelot**

- low-level/basic PDF table extraction
- comes with <u>UI</u> if you don't want to write code
- probably work well for tables with a simple / straightforward structure in computer generated PDFs

## Level 2: cascaded heuristics and ML

- ML and/or heuristics are used to perform structure/layout detection
  - helps making sense complex doc structure and layouts (weird content arrangement, complex tables, diagram, charts...)
- ML is used to read from images (OCR)

### Parsing is "cascaded":

- flow 1:
  - 1. detect doc layout/structure
  - 2. read or OCR each structure item content
- flow 2:
  - 1. read all texts and OCR all images
  - 2. detect doc layout/structure
  - 3. put the text inside the layout boxes bounding boxes

#### Pros:

- works very well on computer generated PDFs
- can understand some complex doc layout/structure
- can read from some images
- well suited for scans and images when used end-to-end

#### Cons:

- slower than just heuristics (ML involved)
- never use both structure and content information at the same time: errors made at one step propagate to the next one
- not the best suited for scans

# docling

- simple, production-ready Python API
- your project, improving very quickly
- supports more input formats than just PDFs (DOCX, XLSX, HTML, images)
- configurable OCR (Tesseract, EasyOCR, macOS OCR, RapidOCR...)
- support enrichments: picture description, classification, formula understanding etc...
- many end-to-end vision models optimized for macOS

# **marker**

- best option to parse tables in computer generated PDFs (as of the end of 2024, might have changed)
- similar to docling, a bit less configurable
- leverages a powerful in-house OCR engine (<u>surya OCR</u>)
- not fully open-source (limitations might apply depending on your org)

# **MinerU**

- similar to marker and docling
- might perform better on Mandarin docs ( open-source project)

#### unstructured

- open-source version of a paid service
- used to have limited features compared to the paid service

# natural-pdf

- "a friendly library for working with PDFs, built on top of pdfplumber" by <u>Jonathan Soma</u>, Data Journalism Professor at Columbia
- detailed walkthroughes and use cases (table extraction, text categorization, question answering...)
- lower level than other alternatives: fine-grained control of PDF structure items
- look at this incredible live demo

### **ABBYY**

- similar to marker / docling / MinerU
- very reliable
- but paid (not open-source)

# Level 3: end-to-end ML parsing

- aka "e2e" or "VLMs"
- leverages both structure and content information to parse a PDF docs into Markdown, HTML, JSON...
- no mode layout detection and OCR, everything is done at once
- computer generated PDFs and scans are equally treated as images

# How is end-to-end parsing helping?

- structure and content are not independent
- understanding structure helps understand content and vice versa

#### **Invoice**



Invoice Number 123

Issue Date March 25, 2024 Due Date March 25, 2024

Status PAID

Example, LLC

123 Fake Street New York City, NY 10012 (555) 867-5309 support@example.com Bill To Customer Address

City, State Zipcode customer@example.org

Item	Unit Cost	Quantity	Amount
Subscription	\$19.00	1	\$19.00
		Subtotal	\$19.00
		Tax Rate	0%
		Amount Due	\$19.00

For questions, contact us anytime at support@example.com.

- if I read a numerical value followed by a "\$" I'm probably reading a table area (content → structure)
- if I'm reading the last row of a table, there's a high chance that numerical can be deduced from the above values (structure content)

# How to parse with end-to-end libs?

- e2e libs go directly from image to Markdown/HTML/JSON
- prompt engineering (uncertain)
- Markdown/HTML/JSON post-processing (easy)

#### Pros:

- yields the best performances on scans and images
- understand complex doc layouts

#### Cons:

- ML hardware acceleration is often required (macOS/GPU)
- sometimes parse when it could "just read" (computer generated PDFs)
- slower and more expensive
- hallucination might become be an issue (depending on the model)

# docling, marker, MinerU

- these libs are actually **hybrid** (level 2 or level 3)
- they are often highly configurable: you can set your tradeoff
   between accuracy and cost/speed by picking the right ML model
- you can also select tiny and specialized models potentially less prone to hallucinations than larger VLMs

# **OlmOCR**

- Mistral OCR like performances (people did report better perf with Mistral OCR or Gemini on their use case)
- "Less \$200 USD per million pages converted"
- you can try it online, it even understands handwritten information

# **GOT-OCR2.0**

• like OlmOCR but with a smaller model

# **Coming soon — Part 2: practice**

# Thank you!