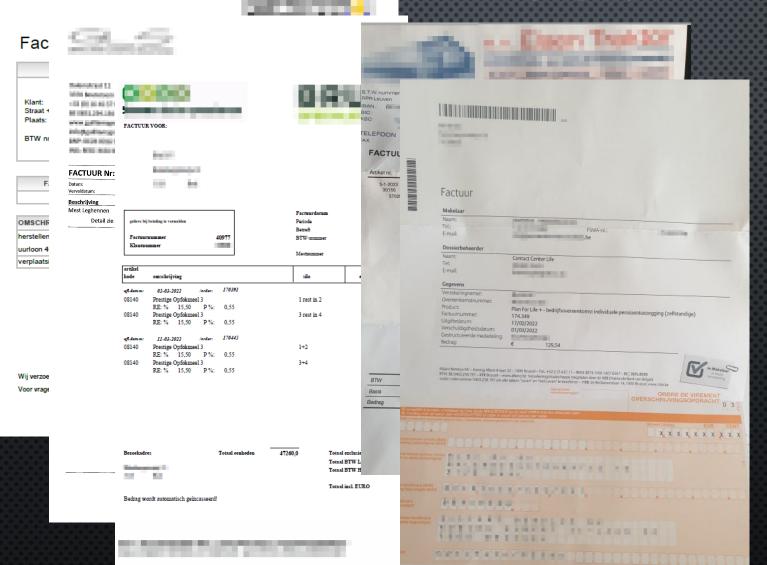
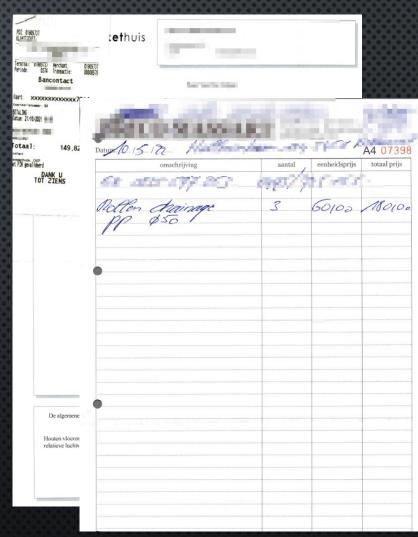
AND LEVERAGING IT WITH A

BY ANDY OEYEN

INVOICES





MANUAL PROCESSING



Invoice types: All kinds

Advantages:
Trained brains
Reliable
Disadvantages:
Slower

STANDARD E-INVOICES PROCESSING

```
<rsm:CrossIndustryInvoice xmlns:rsm="urn:un:unece:uncefact:data:standard:CrossIndustryInvoice:100"</p>
   <ram:GuidelineSpecifiedDocumentContextParameter>
     <ram:ID>urn:cen.eu:en16931:2017</ram:ID>
   </ram:GuidelineSpecifiedDocumentContextParameter>
   <ram:ID>2021 10</ram:ID>
   <ram:TypeCode>380</ram:TypeCode>
    <ram:IssueDateTime>
     <udt:DateTimeString format="102">20210924</udt:DateTimeString>
   </ram:IssueDateTime>
     <ram:AssociatedDocumentLineDocument>
       <ram:LineID>1</ram:LineID>
     </ram:AssociatedDocumentLineDocument>
     <ram:SpecifiedTradeProduct>
       <ram:Name>Project management</ram:Name>
       <ram:Description/>
```

Type of invoices: Digital machine readable invoices with standard format (UBL/ CEFACT/ Zugferd/...)

Method:

Generic algorithm

Advantages:

Very (very) high reliability

Fast

Can be combined with specific algorithms

Disadvantage:

Not commonly used in agriculture (in Belgium)

DIGITAL INVOICE PROCESSING

REO Elect_Factuur_20230714_INCL_contracten.txt

REO Elect_Factuur_20230712_INCL_contracten.txt

REO Elect_Factuur_20230710_INCL_contracten.txt

REO Elect_Factuur_20230707_INCL_contracten.txt

REO Elect_Factuur_20230705_INCL_contracten.txt

REO Elect_Factuur_20230703_INCL_contracten.txt

REO Elect_Factuur_20230630_INCL_contracten.txt

REO Elect_Factuur_20230628_INCL_contracten.txt

REO Elect_Factuur_20230626_INCL_contracten.txt

- Type of invoices:
 Digital machine readable invoices
 XML/JSON/CSV/...
 Method:
 Specific algorithms
- Advantages:
 Very high reliability (depending on how data was generated)
 Fast
- Disadvantage:

 Different algorithm per type

UNSTRUCTURED INVOICE PROCESSING



Type of invoices:

Digital or Digitalised invoice (unstructured)
Paper -> pdf, png, ...

Method:

OCR in combination with specific algorithms and templates

Advantages:

Fast

Disadvantage:

Result depends on quality of input Different algorithm per invoice type Changes



Type of invoices:

Digital or digitalised invoice (unstructured)

Paper -> pdf, png, ...

Method:

Al model

Advantages:

Fast

All kinds of invoice types

Disadvantage:

Accuracy depends on model

Black box is scary

Training takes time

SIDE NOTE

• Most invoice processing methods Still require some kind of mapping to a standard product / product type.

AI PROCESSING: STEP 1

Label Train Predict

AI PROCESSING: STEP 2

Label invoices with low probabilities (Active learning) or known mistakes

Label

Train

Predict

PROOF OF CONCEPT AI IMPLEMENTATION

- Model Microsoft LayoutLMv3
 - NATURAL LANGUAGE PROCESSING
 - COMBINES TEXT WITH LAYOUT
 - Tesseract for OCR
 - Usage: Finetuned the base model

- LABEL SOFTWARE: LABEL-STUDIO
 - CONNECTS TO ML-Albackend
 - ACTIVE LEARNING POSSIBLE

Step 1: Initial prediction with Base model

-> All Labels: OTHER (red)

Step 2: Labels invoices

Step 3: Finetune model

Step 4:
Prediction with finetuned model



RESULTS AND THOUGHTS

- F1 AFTER 1ST ITERATION OF TRAINING WITH 56 LABELLED INVOICES: 0,82
- FURTHER FINETUNING IS NECESSARY
- PREPROCESSING MODEL TO IMPROVE OVERALL ACCURACY
 - HANDWRITTEN INVOICES/ RECEIPTS / DIGITAL INVOICES/SCANNED
- STANDARD E-INVOICES WOULD MAKE THIS OBSOLETE
- OTHER IDEAS ARE WELCOME
- THANK YOU FOR YOUR ATTENTION

MORE INFORMATION:

- HTTPS://GITHUB.COM/MICROSOFT/UNILM/BLOB/MASTER/LAYOUTLMV3/README.MD
- HTTPS://LABELSTUD.IO/