

# Faster DAN: Multi-target Queries with Document Positional Encoding for End-to-end Handwritten Document Recognition

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## Introduction

Handwritten Document Recognition (HDR) = recognize text & layout

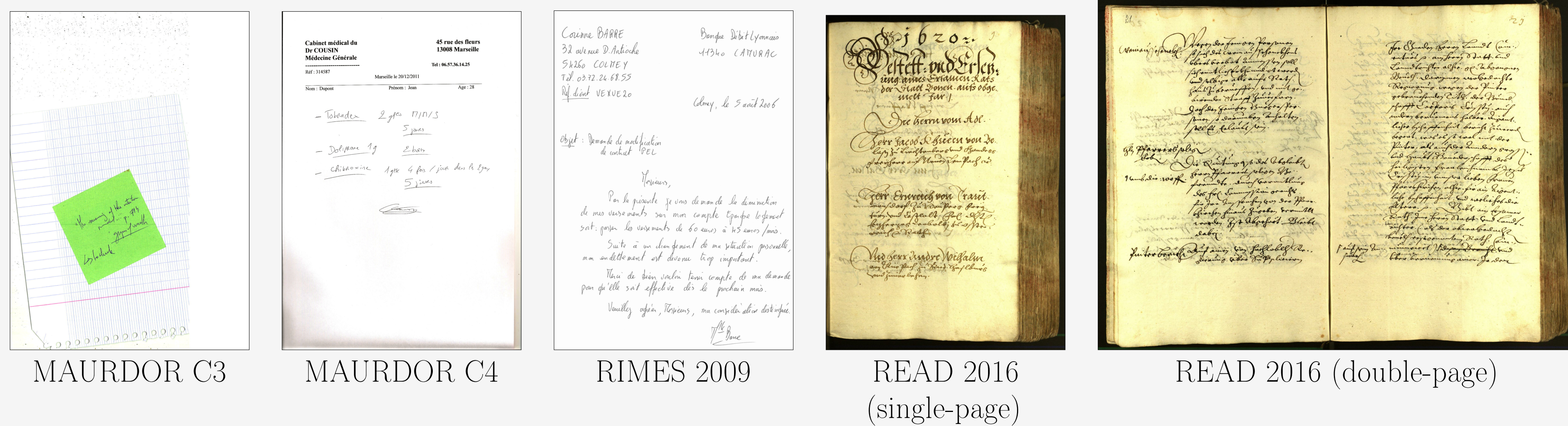
SOTA approach for end-to-end HDR: autoregressive character-level attention decoding process

SOTA model: Document Attention Network (DAN) [1]

► Drawback: prediction time increases with output sequence length (~ 1 second for 100 characters)

**Goal: reducing prediction time**

## Datasets



## Results

| Architecture | READ 2016 (single-page) |        |                      | READ 2016 (double-page) |        |                      | RIMES 2009 (single-page) |        |                      | C3    | C4    |
|--------------|-------------------------|--------|----------------------|-------------------------|--------|----------------------|--------------------------|--------|----------------------|-------|-------|
|              | CER ↓                   | LOER ↓ | mAP <sub>CER</sub> ↑ | CER ↓                   | LOER ↓ | mAP <sub>CER</sub> ↑ | CER ↓                    | LOER ↓ | mAP <sub>CER</sub> ↑ | CER ↓ | CER ↓ |
| DAN [1]      | 3.43                    | 5.17   | 93.32                | 3.70                    | 4.98   | 93.09                | 4.54                     | 3.82   | 93.74                | 8.62  | 8.02  |
| Faster DAN   | 3.95                    | 3.82   | 94.20                | 3.88                    | 3.08   | 94.54                | 6.38                     | 4.48   | 91.00                | 8.93  | 9.88  |

CER: Character Error Rate based on string edit distance to evaluate the text recognition.

LOER: Layout Ordering Error Rate based on graph edit distance to evaluate the layout recognition.

mAP<sub>CER</sub>: mean Average Precision based on a CER threshold to evaluate text & layout recognition altogether.

Prediction time (in seconds, averaged on the test set for a single document image, using a single GPU V100).

|              | RIMES 2009    | READ 2016   |             | MAURDOR |      |         |
|--------------|---------------|-------------|-------------|---------|------|---------|
|              | (single-page) | single-page | double-page | C3      | C4   | C3 & C4 |
| DAN [1]      | 5.6           | 4.6         | 8.5         | 5.8     | 7.7  | 6.6     |
| Faster DAN   | 1.4           | 0.9         | 1.9         | 1.0     | 1.6  | 1.3     |
| Speed factor | x4            | x5.1        | x4.5        | x5.8    | x4.8 | x5.1    |

## Conclusion

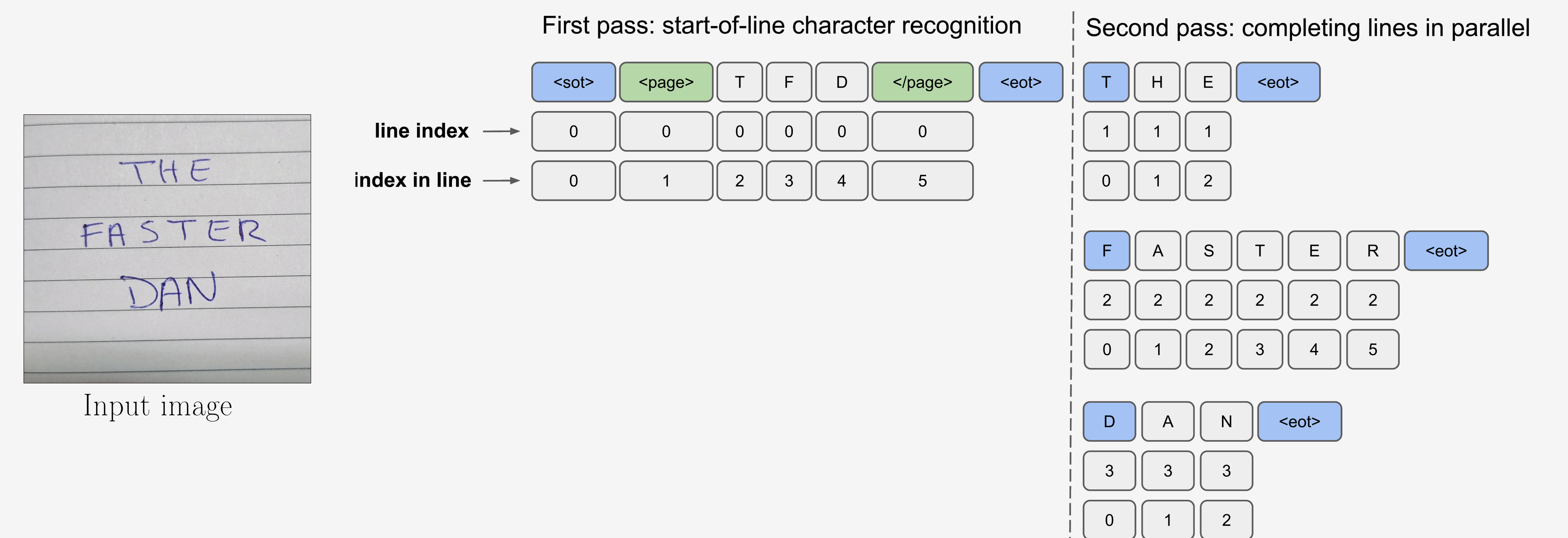
- Generic approach for character-level attention-based models
- Competitive results on three public datasets with the DAN architecture
- At least 4 times faster than sequential prediction process



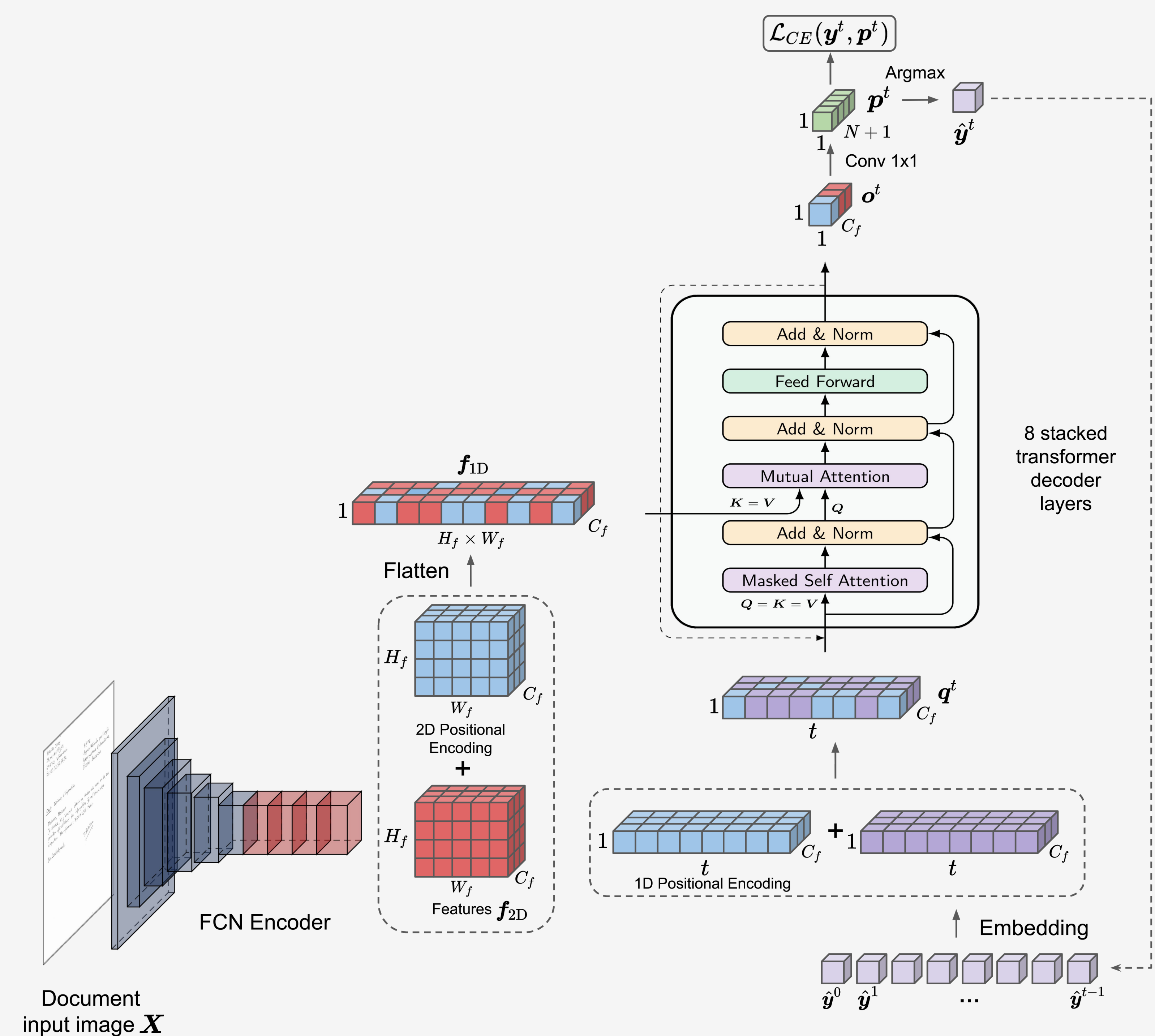
Paper, code and more !

## Approach: parallelizing text line recognition

A two-step decoding process with document positional encoding:



During the first pass, the first character of each line is recognized, as well as the layout tokens (in green). The line index is set to 0.



The DAN architecture

[1] Denis Coquenot, Clément Chatelain, and Thierry Paquet. "DAN: a Segmentation-free Document Attention Network for Handwritten Document Recognition". In: *IEEE Transactions on Pattern Analysis and Machine Intelligence (TPAMI)* 45.7 (2023), pp. 8227–8243. DOI: 10.1109/TPAMI.2023.3235826.