

Attention is all you need to read

Denis Coquenat

2023/09/19

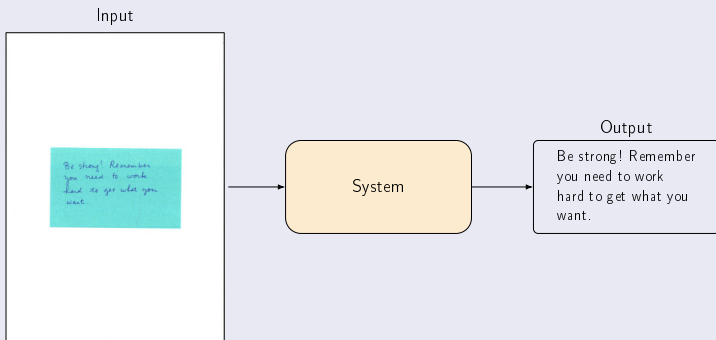


Table of contents

- 1 Context
- 2 Hybrid attention for paragraph recognition
- 3 Autoregressive attention for document recognition
- 4 Speeding up with parallel attention

Handwritten Text Recognition (HTR)

An image-to-sequence problem



Input: an image

Output: a sequence of characters

Challenges

A wide variety of documents

Writing styles, layout, size / resolution, background

to the children any more

but those hopes were dashed.

harvest of which way

when they went to bed

she will undertake a

Challenges

A wide variety of documents

Writing styles, **layout**, size / resolution, background

Wire Transfer Fax Cover Sheet

(Date of Transfer) 10/4/2001 (Transfer Amount) 250

Sender's Information	
Name	Mr. Dupont
Address	12345678901234567890
City, State, Zip	DE 34 5678
Social Security Number	9876543210
Daytime Phone	9876543210
Evening Phone	
Bank Name	Bank of America
Bank Address	12345678901234567890
City, State, Zip	DE 34 5678
Bank Phone Number	9876543210
Routing Number	123456789
Account Number	9876543210

Recipient's Information	
Name	Mrs. Gault
Address	12345678901234567890
City, State, Zip	WA 98 7654
Social Security Number	9876543210
Daytime Phone	9876543210
Evening Phone	
Bank Name and Phone #	Bank of America 9876543210
Bank Address	12345678901234567890
City, State, Zip	WA 98 7654
Bank Phone Number	9876543210
Routing Number	123456789
Account Number	9876543210

Notes:

M. Nicolas Jean Ponce
85 rue d'Or
83 500 Tournillon
03 54 85 08 34

Mabelle Suzanne Sordet
87 rue Pichon
83000 Colomere

Bonjour, à 5 euros 100

Objet : demande de paise en change exceptionnel.

Bonjour Monsieur,

Je vous prie de bien vouloir m'envoyer un chèque exceptionnel de 100 euros, je vous prie de bien vouloir m'envoyer un chèque exceptionnel de mon problème de santé.

En effet, je souffre actuellement d'un état de santé qui m'empêche de travailler. Je vous prie de bien vouloir m'envoyer un chèque exceptionnel de 100 euros, je vous prie de bien vouloir m'envoyer un chèque exceptionnel de mon problème de santé.

C'est pour cette raison que je vous prie de bien vouloir m'envoyer un chèque exceptionnel.

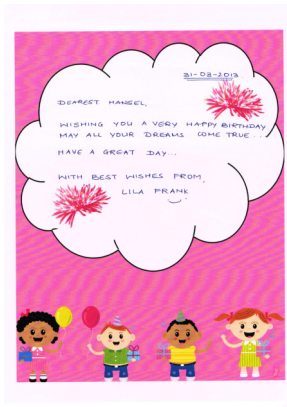
Je vous prie de bien vouloir m'envoyer un chèque exceptionnel de 100 euros, je vous prie de bien vouloir m'envoyer un chèque exceptionnel de mon problème de santé.

M. Jean Ponce Nicolas
Nicolas

Challenges

A wide variety of documents

Writing styles, layout, size / resolution, **background**



Challenges

A wide variety of documents

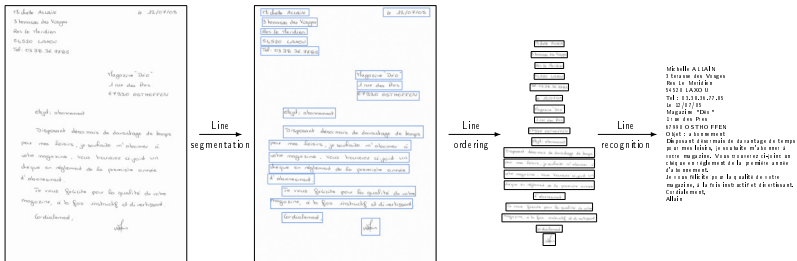
Writing styles, layout, size / resolution, background

No a priori knowledge about the document

- Number of lines
- Number of characters per line
- Reading order

The line-level sequential paradigm

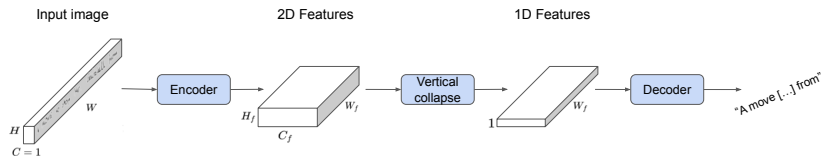
- Segmentation
- Ordering
- Recognition



Related works: Recognition stage

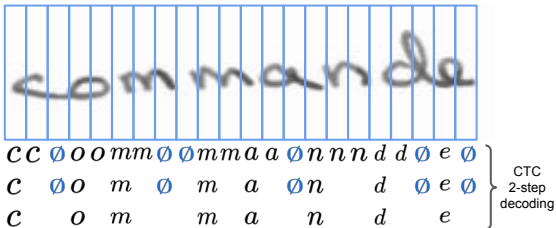
Challenges:

- going from a 2D input image to a 1D sequence of characters
- a variable, unknown number of ordered characters to predict



Related works: Recognition stage

The Connectionist Temporal Classification (CTC) paradigm [1]

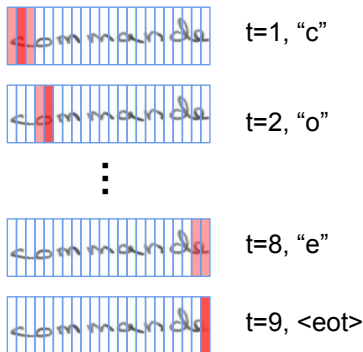


- A frame-by-frame decision process
- Special blank token \emptyset
- A left-to-right constrained alignment
- CTC loss
- Limited to 1d sequences

[1] Graves et al., ICML 2006

Related works: Recognition stage

The attention paradigm (at character level) [2, 3]



- Iterative decoding process
- Implicit character segmentation
- Special end-of-transcription token <eot>
- Unconstrained attention → reading order must be learned
- Cross-Entropy loss

$$c^t = \sum_i \alpha_i^t f_i$$

$$\sum_i \alpha_i^t = 1$$

Conclusion

The sequential paradigm: a mature approach... with some limitations

- Three steps treated independently
- A complex pipeline, hard to maintain
- Cumulative errors between steps
- Additional segmentation annotations
- Rule-based reading order

Goal: to overcome these limitations

Strategy: designing end-to-end HTR models step by step

- ▶ from line to document level

Table of contents

- 1 Context
- 2 Hybrid attention for paragraph recognition
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Related works: Paragraph recognition

Challenges from line to paragraph recognition

- An additional vertical reading order
- Variable number of text lines
- Variable interline spacing, indent

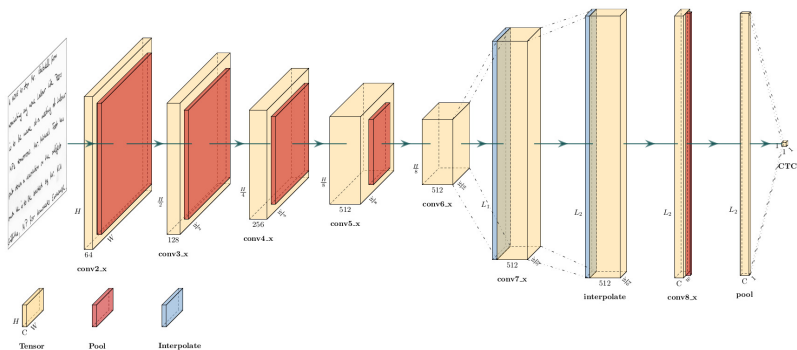
J'ai commandé, il y a une semaine, une paire de chaussette chez vous (n° de réf. client: YZWMLOZ), étant satisfaite de ma commande, je désire en recevoir deux autres paires.

Je vous prie d'agréer Madame, Monsieur, l'expression de nos sentiments distingués.

Related works: Paragraph recognition

CTC-only approaches

- OrigamiNet [4]

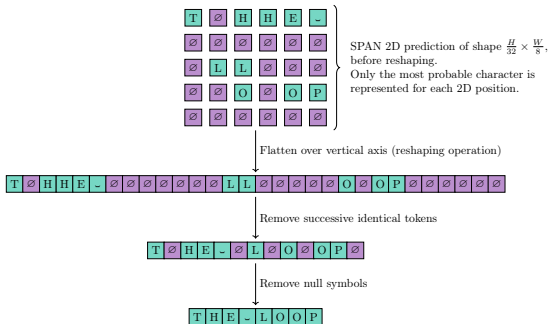


[4] Yousef *et al.*, CVPR 2020

Related works: Paragraph recognition

CTC-only approaches

- OrigamiNet [4]
- **Contribution:** Simple Predict & Align Network (SPAN) [5]



[5] Coquenat *et al.*, ICDAR 2021

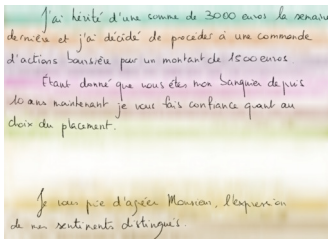
Related works: Paragraph recognition

CTC-only approaches

- OrigamiNet [4]
- **Contribution:** Simple Predict & Align Network (SPAN) [5]

Attention-based approaches

- Line-level attention [6]



[6] Bluche et al., NIPS 2016

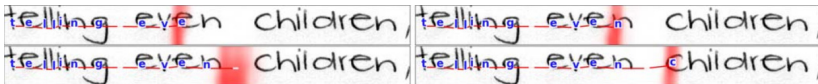
Related works: Paragraph recognition

CTC-only approaches

- OrigamiNet [4]
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Attention-based approaches

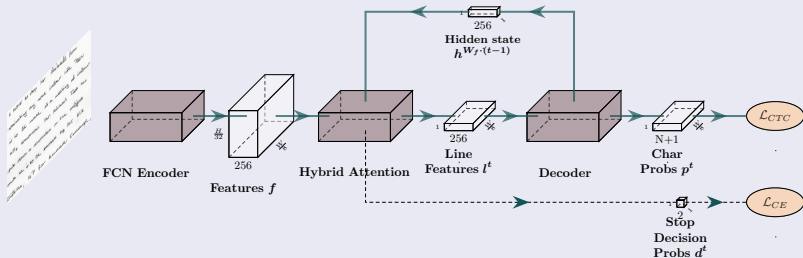
- Line-level attention [6]
- Character-level attention [7, 8]



[7] Bluche *et al.*, ICDAR 2017

Contribution: Vertical Attention Network (VAN) [9]

Overview



[9] Coquenot et al., TPAMI 2023

Line-level vertical hybrid attention

$$\alpha_i^t = \text{softmax}(W_a \tanh(W_f f_i^t + W_j j_{t,i} + W_h h_{W_f(t-1)}))$$

Datasets

Il a bien reçu votre lettre concernant mes anciens
autobus, et m'a promis de les faire
financer. Mais je n'en ai pas encore
vu un seul. Les paiements mensuellement sur une
période de 30 ans.

Je n'ai pas de renseignements de 5 et chaque mois
sont payés par banque, soit par paiement
automatique, à votre convenance. Je n'ai pas
eu une seule fois de 25€ et fin de remboursement.

En espérant que vos anciens fournisseurs de bus
reçoivent, je vous prie d'agréer Madame, l'assurance de
mon respectueux dévouement.

Je vous remercie pour le dossier de ma lettre en matière de
mon compte et dans votre ma nouvelle adresse.

Tout va bien
1 rue d'Alsace
91000 Evry
Avec vous
Avec vous

Je vous remercie de bien vouloir à jour mon dossier et de
me faire connaître vos nouvelles coordonnées.

Je vous en remercie bonne réception.

Je remercie votre comité des dames d'avoir en public
Gustave.

RIMES 2011

This figure has been prepared only on the basis of the
best current information on July 19, 1952. And official
fact it may be too much for the city's safety work.

They will continue and private buses will have to be used.

The main reason considered Mr. WEAVER's
allegation association with organization Street-
led by the Government, immediately Mr.
Wenatchey replied a letter to Senator Tolson
regarding the Federal Bureau of Investigation had
reported on Mr. WEAVER. He believed
the woman perform "outstanding service"
in his past. Senator Tolson's committee
was to pass Mr. WEAVER's nomination before it
can be considered by the full Senate.

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alleged association with organization Street-
led by the Government, immediately Mr.
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Tolson saying the Federal Bureau of In-
vestigation had reported on Mr. WEAVER.
He believed the woman perform "outstanding
service" in his past. Senator Tolson's
committee was to pass Mr. WEAVER's
nomination before it can be con-
sidered by the full Senate.

IAM

Je vous remercie pour le dossier de ma lettre en matière de
mon compte et dans votre ma nouvelle adresse.

Tout va bien
1 rue d'Alsace
91000 Evry
Avec vous
Avec vous

Tout va bien
1 rue d'Alsace
91000 Evry
Avec vous
Avec vous

39

READ 2016

Paragraph-level recognition results

Paragraph-level state-of-the-art approaches, without language model, external data, nor lexicon constraints.

Architecture	IAM		RIMES 2011		READ 2016		# Param.
	CER (%)	WER (%)	CER (%)	WER (%)	CER (%)	WER (%)	
	test	test	test	test	test	test	
Best line-level approach	4.87 ¹		2.3 ²	9.6 ²	4.66 ¹		
[7] CNN+MDLSTM ^b	16.2						
[6] CNN+MDLSTM ^a	7.9	24.6	2.9	12.6			
[8] CNN+Transformer ^b	6.7						27.8 M
[5] SPAN (FCN)	5.45	19.83	4.17	15.61	6.20	25.69	19.2 M
[4] OrigamiNet (GFCN)	4.7						16.4 M
[9] VAN (FCN+LSTM) ^a	4.45	14.55	1.91	6.72	3.59	13.94	2.7 M

¹ Results from [2] CNN+BLSTM^b.

² Results from [10] CNN+BLSTM.

^a With line-level attention.

^b With character-level attention.

VAN demonstration

<https://youtu.be/OXi1birmbuw>

Conclusion

Attention is powerful but:

- Attention mechanisms → slower convergence
 - vertical attention (1D) + pre-training
- Hybrid attention
 - recurrent training (OK for lines, KO for chars)

Bridging the gap between line-level and paragraph-level approaches...

- State-of-the-art results on RIMES 2011, IAM and READ 2016
- Able to deal with slightly inclined lines

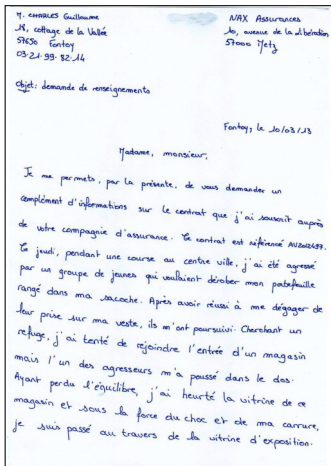
... but still the same limitations, inherent to the sequential paradigm

- Rethinking the paradigm

Table of contents

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HTR at document level

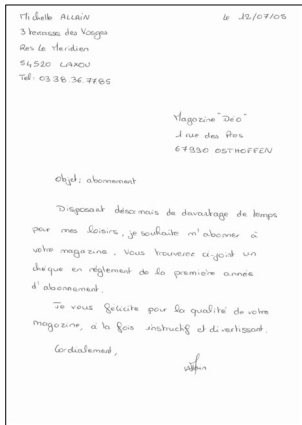


Challenges from paragraph to document

- Layout-dependent reading order
- Larger input images and output sequences
 - GPU constraints
 - More complex attention

Handwritten Document Recognition (HDR)

Goal: joint recognition of both text and layout from whole documents

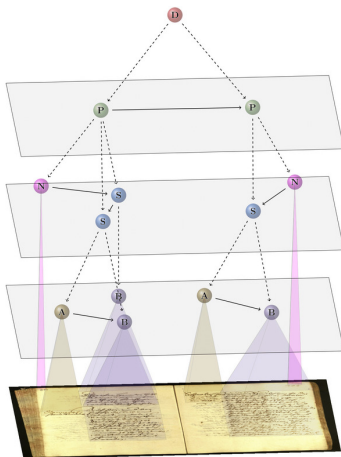


Handwritten Document
→
Recognition

Michelle ALLAIN
3 terrasses des Vosges
Res Le Meridien
54520 LAXOU
Tel : 03.38.36.77.85
Le 12/07/05
Magazine "Déo"
1 rue des Pres
67990 OSTHOFFEN
Objet : abonnement
Disposant désormais de davantage de temps
pour mes loisirs, je souhaite m'abonner à
votre magazine. Vous trouverez ci-joint un
chèque en règlement de la première année
d'abonnement.
Je vous félicite pour la qualité de votre
magazine, à la fois instructif et divertissant.
Cordialement.
Allain

Sender Coordinates
Recipient Coordinates
Place & Date
Object
Body
Signature

How to encode both text and layout ?



```

<document>
  <page>
    <page_number>
      204
    </page_number>
    <section>
      <body>
        Schgrafer, [...] gehalt.
      </body>
    </section>
    <section>
      <annotation>
        General [...] Raitung
      </annotation>
      <body>
        Auf den: [...] werden,
      </body>
    </section>
  </page>
</page>
<page>
  <page_number>
    204
  </page_number>
  <section>
    <annotation>
      Schmalz. [...] bet:
    </annotation>
    <body>
      Verer [...] dar
    </body>
  </section>
</page>
</document>

```

► XML paradigm

How to evaluate the performance ?

Evaluate the text recognition

- CER / WER
- ▶ Normalized edit distance between sequences of characters / words

Prediction: "<A>HTR2HDR"

Metric computed on: "HTR2HDR"

How to evaluate the performance ?

Evaluate the text recognition

- CER / WER

Evaluate the layout recognition

- LOER (Layout Ordering Error Rate)
- ▶ Normalized edit distance between graphs

Prediction: "`<A>HTR2HDR`"

Metric computed on: "`<A>`"

How to evaluate the performance ?

Evaluate the text recognition

- CER / WER

Evaluate the layout recognition

- LOER (Layout Ordering Error Rate)

⚠ **Not sufficient:**

Ground truth: "`<A>HTR2HDR`"

Prediction: "`<A>HTR2HDR`"

LOER = 0% CER = 0%

How to evaluate the performance ?

Evaluate the text recognition

- CER / WER

Evaluate the layout recognition

- LOER (Layout Ordering Error Rate)

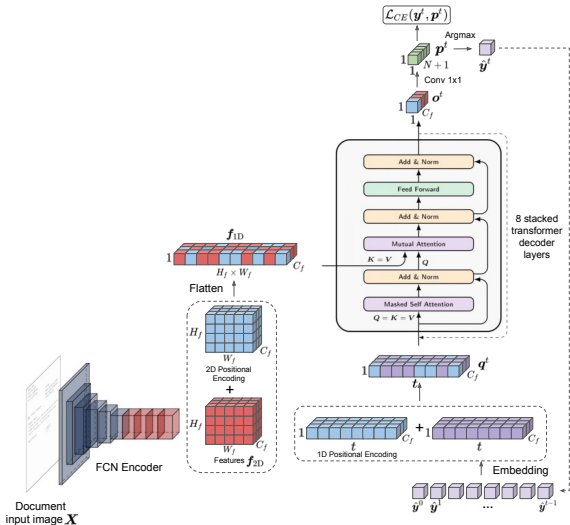
Evaluate text and layout recognition altogether

- mAP_{CER}
- ▶ Area under the precision / recall curve

Prediction: "**<A>**HTR****2****HDR********"

Metric computed on: "**HTR**2**HDR**", "**HTR**", "**HDR**"

Document Attention Network (DAN) [11]



$$\mathbf{c}^t = \underbrace{\text{softmax}\left(\frac{QK^T}{\sqrt{d_k}}\right)}_{\alpha^t} \mathbf{V}$$

→ Teacher forcing

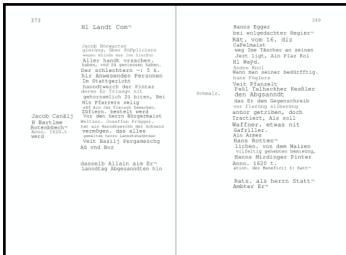
[11] Coquet et al., TPAMI 2023

DAN - Training strategy

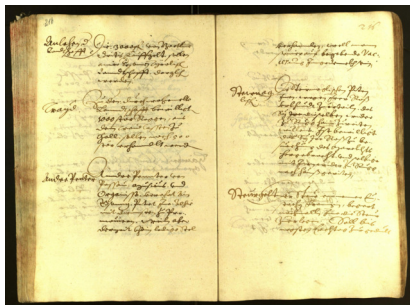
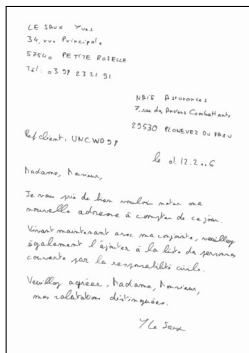
- Pre-training on synthetic text line images.
- Curriculum learning with synthetic documents:

(a) $l = 3$.

957	First die die Organisten Zuziehen sich mit die Firscheid, dunt in Ecken Vlast, Lied und Vast, der die die die die die Bericht, die die nach in einen von And weiden. Nulgepp. Luden erhalten, von Dixxy Flinter, Hörwarter die, durch	Es referieren, Nulle hoffen bewilligt.	936 In sagen Amlichen der, Pias, als einen Fall Leben Zücker, In legen der, Aussetzung meer- bewilligt, Im Pflaster, werden, man will worthe drei Tag Termin Jacob Gabel dable) es sich kleibt, Hans Fingletter
-----	---	---	--

(b) $l = 15$.(c) $l = l_{\max} = 30$ (end of curriculum stage, no crop).

Datasets



Dataset	Level	Training	Validation	Test	# char tokens	# layout tokens
RIMES 2009 [12]	Page	1,050	100	100	108	14
READ 2016 [13]	Page	350	50	50	89	10
	Double page	169	24	24		

DAN results on the RIMES dataset

⚠ Metrics do not take into account the segmentation step

Dataset	Approach	CER (%) ↓	WER (%) ↓	LOER (%) ↓	mAP _{CER} (%) ↑
RIMES 2011	Line level				
	[9] FCN	3.04	8.32	✗	✗
	[10] CNN+BLSTM ^a	2.3	9.6	✗	✗
	[11] DAN (FCN+transformer) ^c	2.63	6.78	✗	✗
	Paragraph level				
	[5] SPAN (FCN)	4.17	15.61	✗	✗
	[6] CNN+MDLSTM ^b	2.9	12.6	✗	✗
	[9] VAN (FCN+LSTM) ^b	1.91	6.72	✗	✗
[11] DAN (FCN+transformer) ^c	1.82	5.03	✗	✗	
RIMES 2009	Paragraph level				
	[11] DAN (FCN+transformer) ^c	5.46	13.04	✗	✗
	Page level				
[11] DAN (FCN+transformer) ^c	4.54	11.85	3.82	93.74	

^a This work uses a slightly different split (10,203 for training, 1,130 for validation and 778 for test).

^b with line-level attention.

^c with character-level attention.

DAN results on the READ 2016 dataset

⚠ Metrics do not take into account the segmentation step

Approach	CER (%) ↓	WER (%) ↓	LOER (%) ↓	mAP _{CER} (%) ↑
Line level				
[2] CNN+BLSTM ^a	4.66	X	X	X
[13] CNN+RNN	5.1	21.1	X	X
[9] VAN (FCN+LSTM) ^b	4.10	16.29	X	X
[11] DAN (FCN+transformer) ^a	4.10	17.64	X	X
Paragraph level				
[5] SPAN (FCN)	6.20	25.69	X	X
[9] VAN (FCN+LSTM) ^b	3.59	13.94	X	X
[11] DAN (FCN+transformer) ^a	3.22	13.63	X	X
Single-page level				
[11] DAN (FCN+transformer) ^a	3.53	13.33	5.94	92.57
Double-page level				
[11] DAN (FCN+transformer) ^a	3.69	14.20	4.60	93.92

^a with character-level attention.

^b with line-level attention.

DAN demonstration

<https://youtu.be/HrrUsQfW66E>

Conclusion

DAN: the first end-to-end model for HDR

- ▶ Structured output sequence
- ▶ No need for any physical segmentation annotation
- ▶ Can follow the slant of the lines (character-level attention)

Line-level / paragraph-level limitations

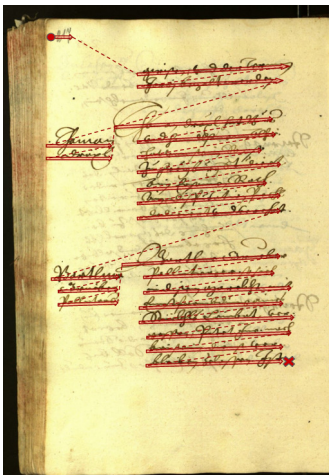
- ~~Three steps treated independently~~
- ~~A complex pipeline, hard to maintain~~
- ~~Cumulative errors between steps~~
- ~~Additional segmentation annotations~~
- ~~Rule-based reading order~~

Drawback: prediction times grow with the character sequence

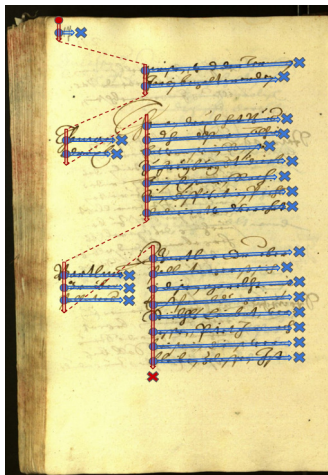
Table of contents

- 1 Context
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Faster DAN: parallelizing text line recognition [14]



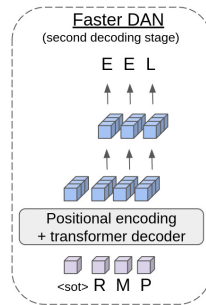
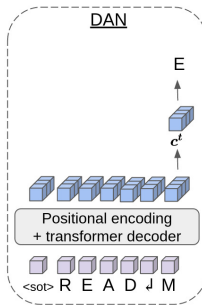
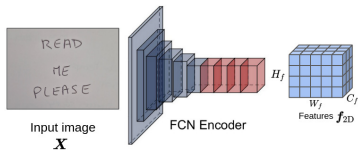
(a) DAN



(b) Faster DAN

[14] Coquenot et al., ICDAR 2023

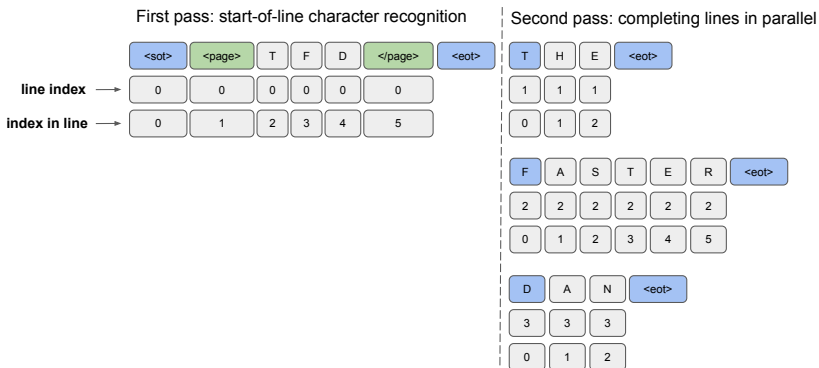
Faster DAN - Multi-target queries



Faster DAN - Positional encoding

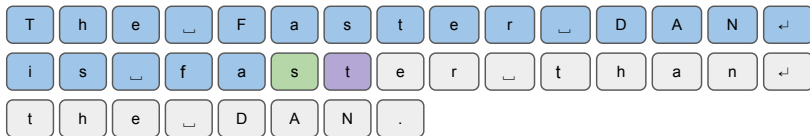


(a) DAN single-pass prediction process



(b) Faster DAN two-pass prediction process

Faster DAN - Context



(a) Context used by the DAN



(b) Context used by the Faster DAN

Results

Architecture	READ 2016 (single-page)				READ 2016 (double-page)			
	CER ↓	WER ↓	LOER ↓	mAP _{CER} ↑	CER ↓	WER ↓	LOER ↓	mAP _{CER} ↑
DAN [11]	3.43	13.05	5.17	93.32	3.70	14.15	4.98	93.09
Faster DAN [14]	3.95	14.06	3.82	94.20	3.88	14.97	3.08	94.54

Architecture	RIMES 2009			
	CER ↓	WER ↓	LOER ↓	mAP _{CER} ↑
DAN [11]	4.54	11.85	3.82	93.74
Faster DAN [14]	6.38	13.69	4.48	91.00

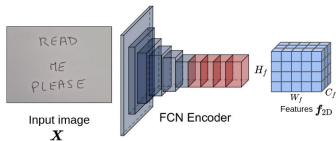
Prediction times

	RIMES 2009	READ 2016		MAURDOR		
		single-page	double-page	C3	C4	C3 & C4
Dataset details (averaged for a document on the test set)						
width (px)	1,235	1,190	2,380	1,336	1,240	1,297
height (px)	1,751	1,755	1,755	1,658	1,754	1,697
# chars	578	528	1,062	481	706	575
# lines	18	23	47	16	22	18
# chars / line	31	22	22	30	31	30
# layout tokens	11	15	30	0	0	0
Prediction times (in seconds)						
DAN [11]	5.6	4.6	8.5	5.8	7.7	6.6
Faster DAN [14]	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

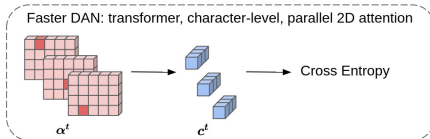
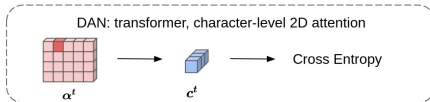
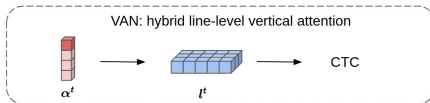
Faster DAN demonstration

https://youtu.be/_pBs02W8XRE

VAN vs DAN vs Faster DAN



← One-shot encoding →



← Iterative decoding →

General conclusion

Attention for reading systems

Line → Paragraph → Document

► From text recognition to reading

Perspectives

Recognizing more:

- Heterogeneous documents (layout)
- Multilingual documents
- Combining HDR with other tasks: Named Entity Recognition, Mathematical Expression Recognition, Table Recognition

Thank you for your attention

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