End-to-end Handwritten Paragraph Text Recognition Using a Vertical Attention Network

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Context







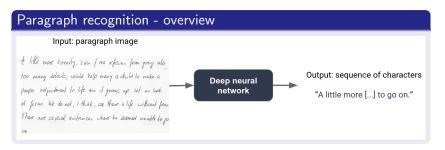




Handwritten text recognition

Task

- Goal: to recognize the text from a scanned document, in the correct order
- Constraints: writing diversity (character shape, slant, size, spacing), input of variable sizes (number of pixels, characters, words, lines), arbitrary document layout with strong variabilities



State of the art

Explicit line segmentation, followed by text line recognition

- Segmentation based on object detection approach: [1], [2]
- Segmentation based on start-of-line prediction: [3], [4]
- → Needs line-level segmentation label and cumulates errors

End-to-end implicit segmentation & recognition

- Attention-based models: line-level [5], character-level [6]
- FCN models: [7], [8]

Datasets

Characteristics

Dataset	Level	Training	Validation	Test	charset size	Language	# lines
RIMES [9]	Line	10 532	801	778	100	French	
	Paragraph	1 400	100	100	100	French	2-18
IAM [10]	Line	6 482	976	2 915	79	English	
	Paragraph	747	116	336	19	Liigiisii	2-13
READ 2016 [11]	Line	8 349	1 040	1 138	89	Early Modern	
	Paragraph	1 584	179	197	09	German	1-26

- Handwritings
- Resolution of 300 dpi

Datasets

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RIMES

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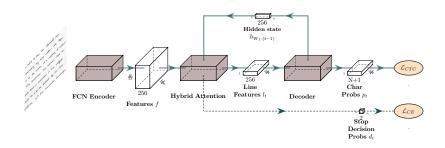
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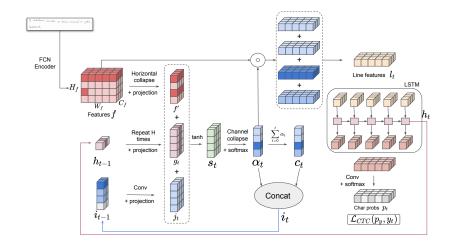


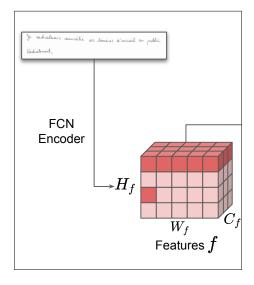
READ 2016

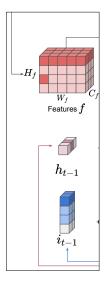
VAN: Vertical Attention Network [12]



• Seg2seg architecture with attention (recurrent process)





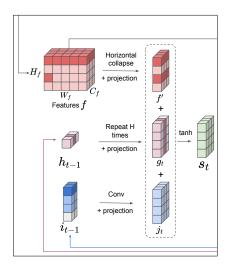


Hybrid attention:

 \rightarrow location-based + content-based

Dimension mismatching:

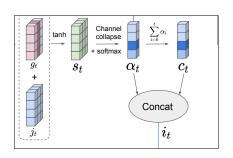
- Features $f \in \mathbb{R}^{H_f \times W_f \times C_f}$
- Decoder LSTM hidden state $h_{t-1} \in \mathbb{R}^{1 \times 1 \times C_h}$
- Information from previous attention weights $i_{t-1} \in \mathbb{R}^{H_f \times 1 \times 2}$



$$s_{t,i} = \tanh(f_i' + g_{t,i} + j_{t,i})$$

Datasets

VAN - Focus on the hybrid attention mechanism

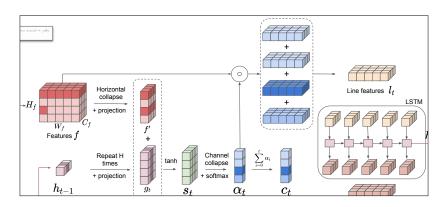


Attention score: $e_{t,i} = W_a \cdot s_{t,i}$

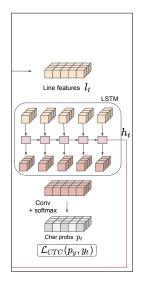
Attention weights:

$$\alpha_{t,i} = \frac{\exp(e_{t,i})}{\sum_{k=1}^{H_f} \exp(e_{t,k})}$$

Coverage vector:
$$c_t = \sum_{k=0}^t \alpha_k$$



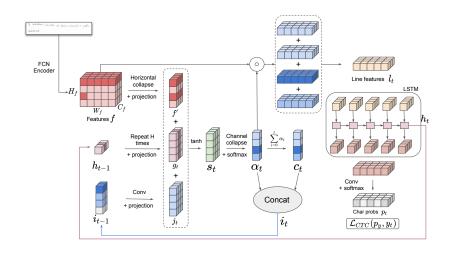
$$I_t = \sum_{i=1}^{H_f} \alpha_{t,i} \cdot f_i$$



Line-by-line alignment with the CTC loss:

$$\mathcal{L} = \sum_{k=1}^{L} \mathcal{L}_{\text{CTC}}(p_k, y_k)$$

where L is the number of lines.



VAN - Visualization



Je me permet de vous écrire ca**s** je v**M**ux



en commander 100.



augmente**s** mes quantités de CD vierges



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New Jarraca

j'ai commandé 50 CD et aduellement je voudrai

Results

Architecture	A+++:	IAM		RIMES		READ 2016		" D
Architecture	Attention	CER (%)	WER (%)	CER (%)	WER (%)	CER (%)	WER (%)	# Param.
Two-step approaches								
[1] RPN+CNN+BLSTM	×	15.6						
[2] RPN+CNN+BLSTM	X	8.5						
[13] RPN+CNN+BLSTM	×	6.4	23.2	2.1	9.3			
End-to-end approaches								
[7] FCN	X	4.7						16.4 M
[8] FCN	×	5.45	19.83	4.17	15.61	6.20	25.69	19.2 M
[6] CNN+MDLSTM	character	16.2						
[14] CNN+Transformer	character	6.7						27 M
[5] CNN+MDLSTM	line	7.9	24.6	2.9	12.6			
[12] Ours (VAN) - FCN+LSTM	line	4.45	14.55	1.91	6.72	3.59	13.94	2.7 M

Conclusion

VAN

- Seq2seq architecture using hybrid attention
- Results beyond the state of the art on three datasets

Limitations

- Pretraining requirements (isolated lines / cross dataset)
- Limited to single-column document

Paper

- Under review (Arxiv: "End-to-end Handwritten Paragraph Text Recognition Using a Vertical Attention Network")
- Available source code and pretrained weights: https://github.com/FactoDeepLearning/VerticalAttentionOCR

Perspective

Handle complex layout

- Multi-column of texts, non-textual items...
- Use transformer architecture with attention at character level
 [15, 14]

New challenges

- Lack of public datasets
- Images of whole documents more voluminous
- Doubtful reading order (maps, schema)
 - \rightarrow inadequate metrics / losses
- From line to character attention
 - \rightarrow the number of iterations increases dramatically

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- Manuel Carbonell et al. "End-to-End Handwritten Text Detection and Transcription in Full Pages". In: Workshop on Machine Learning, WML@ICDAR. 2019, pp. 29–34.
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Reference III

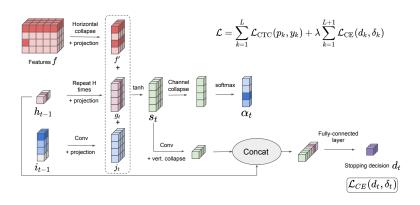
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- [13] Curtis Wigington et al. "Start, Follow, Read: End-to-End Full-Page Handwriting Recognition". In: European Conference on Computer Vision. Vol. 11210. Lecture Notes in Computer Science. 2018, pp. 372–388.
- [14] Sumeet S. Singh and Sergey Karayev. "Full Page Handwriting Recognition via Image to Sequence Extraction". In: <u>CoRR</u> (2021). URL: https://arxiv.org/abs/2103.06450.

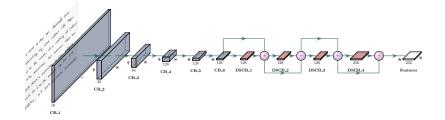
Reference IV

[15] Ashish Vaswani et al. "Attention is All you Need". In: Annual Conference on Neural Information Processing Systems. 2017, pp. 5998–6008.

VAN - End-of-paragraph detection

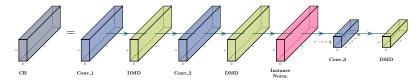


VAN - FCN Encoder



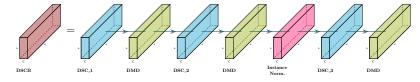
CB: Convolution Block, DSC: Depthwise Separable Convolution Block

CB: Convolution Block



DMD: Diffused Mix Dropout

DSCB : Depthwise Separable Convolution Block



DMD: Diffused Mix Dropout

Training details

- Preprocessing: resolution reduced to 150 dpi + normalization
- Data augmentation: contrast, lightness, resolution, perspective, projection, elastic distortion, dilation and erosion
- Optimizer : Adam
- Initial learning rate: 10^{-4}
- Mini-batch size: 16 for lines, 8 for the VAN

CTC



Connectionist Temporal Classification

Image from Hannun, "Sequence Modeling with CTC", Distill, 2017.