



Watch Your Strokes: Improving Handwritten Text Recognition with Deformable Convolutions



Iulian Cojocaru, Silvia Cascianelli, Lorenzo Baraldi, Massimiliano Corsini, Rita Cucchiara





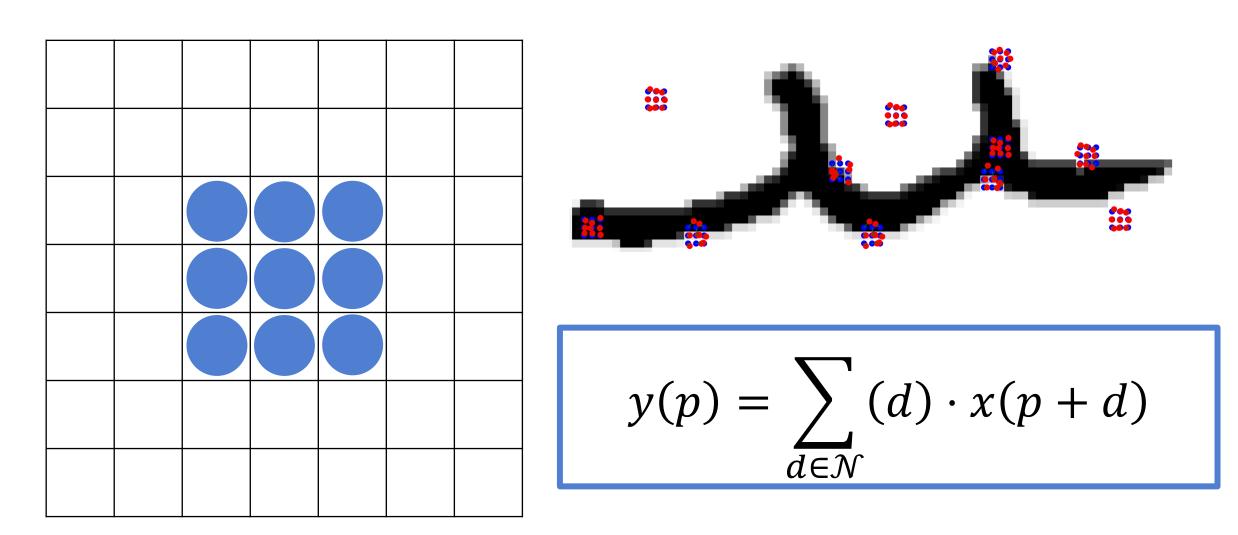
OVERVIEW

Handwritten text is a sparse structure, and characters and words vary in shape, scale, and orientation

→ We take into account these characteristics by using DefConvs in place of standard convolutions for HTR

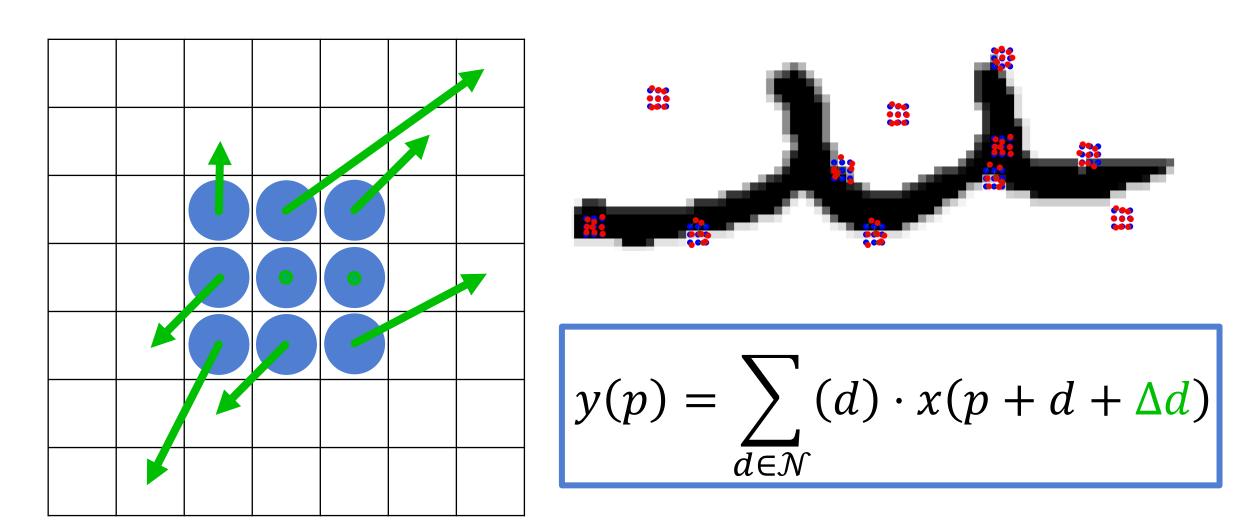


DEFCONVS





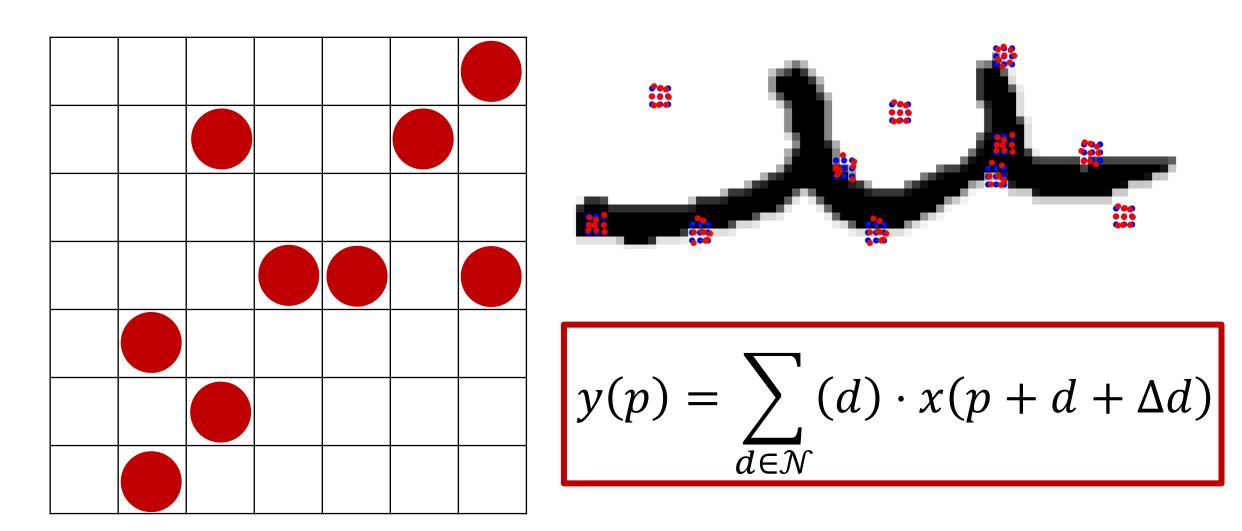
DEFCONVS



[1] J. Dai, H. Qi, Y. Xiong, Y. Li, G. Zhang, H. Hu, and Y. Wei. Deformable convolutional networks. In CVPR, 2017



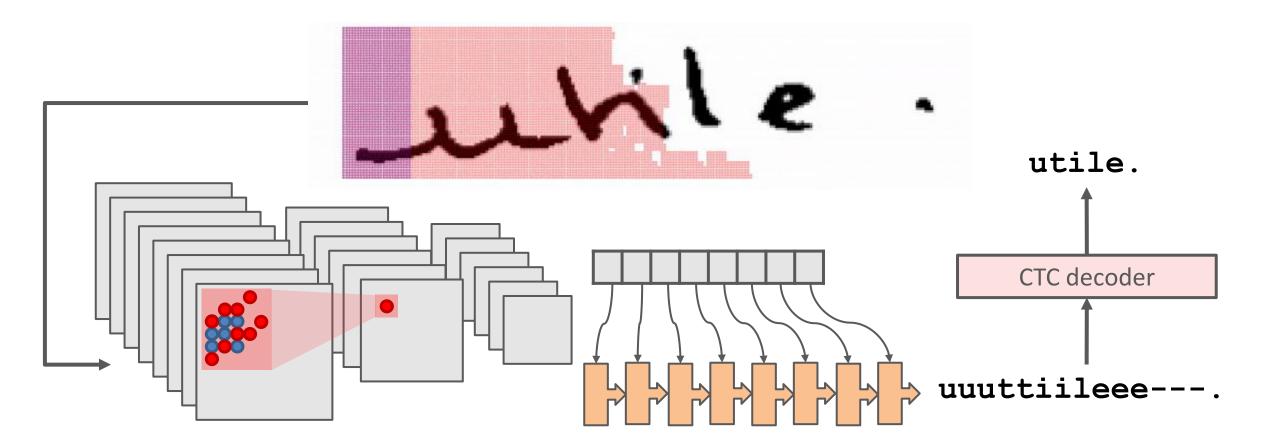
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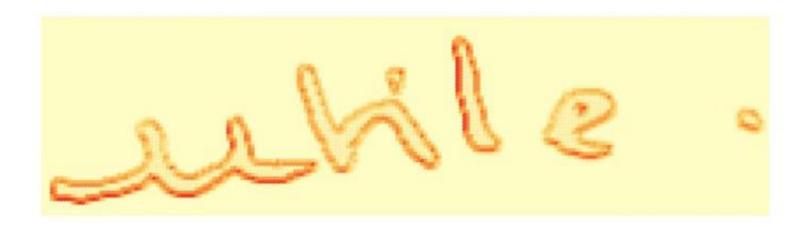




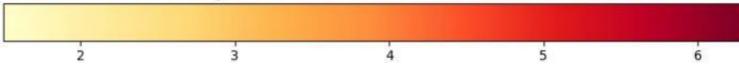
[2] B. Shi, X. Bai, and C. Yao. An end-to-end trainable neural network for image-based sequence recognition and its application to scene text recognition. IEEE Trans. PAMI, 39:2298–2304, 2016.



The falle fixed of the family, love the



Cumulative offsets magnitude





	IAM d	lataset	RIMES dataset		
	CER	WER	CER	WER	
Full-DefConv	4,6	19,3	4,6	14,8	
Shi <i>et al</i> . [2]	5,7	23,2	5,3	17,5	
Wigington <i>et al</i> . [3]	6,4	23,2	2,1	9,3	
Voigtleander <i>et al.</i> [4] – LM	8,3	27,5	4,0	17,7	
Puigcerver [5]	6,2	20,2	2,6	10,7	
Bluche [6]	7,9	24,6	2,9	12,6	
Pham <i>et al.</i> [7]	10,8	32,1	6,8	28,5	

[2] B. Shi, X. Bai, and C. Yao. An end-to-end trainable neural network for image-based sequence recognition and its application to scene text recognition. IEEE Trans. PAMI, 39:2298–2304, 2016.

[3] C. Wigington, C. Tensmeyer, B. Davis, W. Barrett, B. Price, and S. Co-hen. Start, follow, read: End-to-end full-page handwriting recognition. In ECCV, 2018.

[4] P. Voigtlaender, P. Doetsch, and H. Ney. Handwriting recognition with large multidimensional long short-term memory recurrent neural net-works. In ICFHR, 2016.

[5] J. Puigcerver. Are multidimensional recurrent layers really necessary for handwritten text recognition? In ICDAR, 2017.

[6] T. Bluche. Joint line segmentation and transcription for end-to-end hand-written paragraph recognition. In NeurIPS, 2016.

[7] V. Pham, T. Bluche, C. Kermorvant, and J. Louradour. Dropout improves recurrent neural networks for handwriting recognition. In ICFHR, 2014.



AImage^{Lab}

On the IAM dataset:

did not act as though he found it necessary

Groud Truth: Shi et al. [2]:

did not act as though he found it necessary Full-DefConv: did not act as though he found it necessay dd n act as thaugh kefanod it necarseay

latte had lost it life-temps, as the heart

Groud Truth: Full-DefConv: Shi *et al.* in [2]:

earth had lost its life-tempo, as the heart earth had lost its lefe-tempo, as the heart earthled bost its eferteupo, as the beat

liked to go where during his off-duty periods

Groud Truth: liked during his off-duty periods Full-DefConv: liked # during his off-duty periods Shi et al. [2]: liked tegotere during his off-duty periots

On the RIMES dataset:

11ª Dabois, se souhaitersis stre consut ou titre de la responsabilité

Groud Truth:	Md Dubois je souhaiterais être couvert au titr			
	de la responsabilité			
Full-DefConv:	Ma Duois je souhaiterais être coupet au. Titre			
	de l ressonsabilité			
Shi <i>et al.</i> in[2]:	ma Bubois. Je souhatersir être lea mert u titre			
	de b ressonssbilité			
TI ME DEPART				

JE ME PERMET DE VOUS ECRIRE POUR AVOIR

Groud Truth:	JE ME PERMETS DE VOUS ECRIRE POUR AVOIR
Full-DefConv:	JE ME MERMETS DE VOUS ECPIR POUP QNOIR
Shi <i>et al.</i> in [2]:	JEe PERMET DE voUS FcAIrRe Pour avoin

merci de votre collaboration

Groud Truth: merci de votre collaboration Full-DefConv: erci de votre collaloration Shi et al. in [2]: A'acexio de votre collaloration



	IAM dataset				RIMES dataset			
	Full-DefConv		Shi <i>et al</i> . [2]		Full-DefConv		Shi <i>et al</i> . [2]	
	CER	WER	CER	WER	CER	WER	CER	WER
g(0,10) while .	4,7	19,5	5,8	23,7	4,6	14,8	5,3	17,3
9(0,20) uhle.	5,5	22,2	6,9	26,5	4,7	15,4	5,4	18,2
§(0,30) uhle.	8,3	49,0	24,4	62,8	5,1	17,0	6,0	20,2
9(0,10) uhle.	4,8	19,8	5,9	24,0	4,6	14,8	5,3	17,4
9(0,20) uhle.	5,5	22,0	6,7	26,0	4,6	15,1	5,4	17,7
9(0,30) while .	10,6	33,3	13,6	41,2	4,7	15,1	5,5	18,2

[2] B. Shi, X. Bai, and C. Yao. An end-to-end trainable neural network for image-based sequence recognition and its application to scene text recognition. IEEE Trans. PAMI, 39:2298–2304, 2016.



We showed that DefConvs are more suitable than standard convolutions for the

task HTR task since they

- Can adapt to highly distorted handwritten strokes
- Are robust to background noise

→ These features make them promising for being applied for HTR of historical manuscripts (both from benchmark datasets and new ones that we are currently collecting)







Thank you!

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For questions and discussions, please "drop by" our poster or reach out to Silvia Cascianelli: silvia.cascianelli@unimore.it