

Problem description & pipe line Photo CCR pipeline 1) Text detection buck down text 1) Text detection into characters 2) Character seg muntation. 3) Character olassification Through lext detection character segmentation may involve machine learning Feer pedestrian detection in photos (hu aspect ration of a person tends to be almost the same) They to well note almost the same) To sear leads an algo pedestrian detection get a date set of the same of the same pixel size -> supervised learning (87 x 26 paths suppose) Tictory			
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	16 tear teach an algo pedestrian detect	son	
	medadowns Images from the come	pixal	
	Size -> Supervised learning 607 × 26	patches surrose	
			4

<i>y</i>	Date:
*	Suppose ne have a big image 8 ne finish
7	to find pedestrians in it via the pedestrian
	classifier
*	(Mole: there can be multiple gredestrians in
*	lest image in any location)
79	
19	Strding windon detection - take 82 × 36
*	patches of the test image by stiding over
*	The image a few pixels at a time 8
•	checking if pedestrian is present there
•	
-	In The stiding of the pixels the # of
-9	pixels slided over is called
•	step size/strille
-	
-	7 Then therease Palestage image patch size
	8 detect for de pedestrians
	to you will have to convert the larger
ALL V	Inage into 82×36 for the classifier
	I last delection the aspect who as the
	In text detection the aspect vatio of the
-	best 15 not the same. Once sliding window
	detection is done then apply "expansion" operator to results -> expands regions where
-	
**************************************	text was defected to form a rectangle.
	is how we combat the
1	Victory different aspect valor problem Page No.

Date
Character segmentation -
Thain such a classifier where give examples
denotes a portion of the overall best
where there is a spt gap between 2 characters,
-re where this does not exist.
Once trained, then do ID stilling window
to figure out where different characters are
By placing boundaries between letters
(boundary right)
(in one middle)
Cretting lots of data: periorks in certain of situations only
1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
Artificial data Synthesis
Artificial data synthesis for photo CCR
There are many word processing softwares with
There are many word processing softwares with many different types of fonts. You hose forms
I paste on random background to generate
artificial data make data from ground up
You can also take real data & introduce
distortions

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Scanned by CamScanner

				Date:			
Distortion	introduced of n	should	be	reps	esenla	ti'on	of
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