

# C<sup>2</sup>TD: a fast and robust text detector via centre line regression and centre border probability

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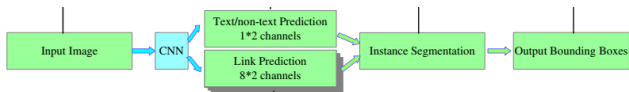
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- 1 Existing problems
- 2 Inspiration
- 3 C<sup>2</sup>TD
- 4 Dewarp Text Lines

## Heavy Computation hampers real word use

- pixel-link: densely predicting link directions, which leads to a complex post-processing.



- PSENet: predicting score maps multiple times.

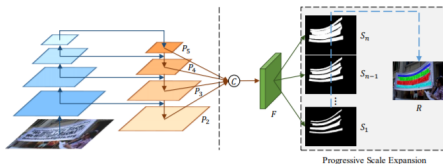


Figure 2: Illustration of our overall pipeline. The left part is implemented from FPN [16]. The right part denotes the feature fusion and the progressive scale expansion algorithm.

Link directions are used to separate different instances. The scale expansion in PSENet has similar effects. Do we have to do it this way?

# predicting text line throughout a image

Some methods (TextBox++, East, etc. al) are not suitable for predicting extremely long texts in nature.

- East: A fast anchor-free method.

- Textboxes: modify the shape of conv kernels & anchor boxes

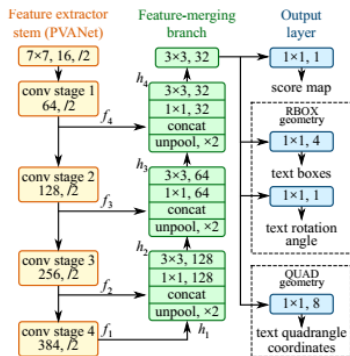
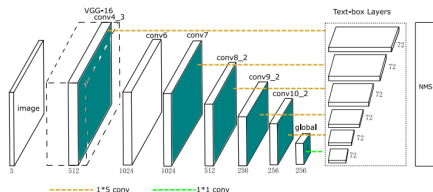


Figure 3. Structure of our text detection FCN.



This detection-based methods run fast, but may only suitable for predicting language that can be separate into words easily.

Only predicting key points(head & tail of a bounding box) is not robust, especially when data contain too much noise.

- Advanced East: predicting two key points. when the key points are not correctly predicted, things get complicated.



What is the problems? give your ideas.

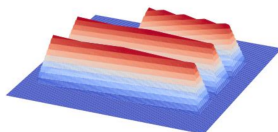
内部像素?	边界像素?	头(or尾)?	左上(右上)X	左上(右上)Y	左下(右下)X	左下(右下)Y
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1. 将叶面积相等的甲、乙两种植物的叶分别放在相同温度、湿度适宜的密闭小室中，  
2. 在充足的光照下，甲、乙两种植物叶片的CO<sub>2</sub>吸收速率分别为10μmol/m<sup>2</sup>/s和5μmol/m<sup>2</sup>/s，  
3. 甲、乙两种植物叶片的CO<sub>2</sub>吸收速率分别为10μmol/m<sup>2</sup>/s和5μmol/m<sup>2</sup>/s，  
4. 甲、乙两种植物叶片的CO<sub>2</sub>吸收速率分别为10μmol/m<sup>2</sup>/s和5μmol/m<sup>2</sup>/s，  
5. 若实验一段时间后，甲叶片所在小室中的CO<sub>2</sub>浓度较初始低，则甲固定CO<sub>2</sub>的能力较低。

Figure: Activation pixels of advanced East

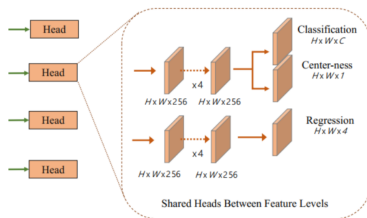
# Inspiration: TextMountain & FCOS

- TextMountain: claim that *the mountaintop can separate text instances which cannot be easily achieved using semantic segmentation map*



Center-border  
Probability

- FCOS: Fully Convolutional One-Stage Object Detection.



$$centerness = \sqrt{\frac{\min(l, r)}{\max(l, r)} * \frac{\min(t, b)}{\max(t, b)}} \quad (1)$$

when  $l = r$  and  $t = b$ , the centerness reaches extremum.

FCOS was accepted by CVPR 2019

# C<sup>2</sup>TD: a fast and robust text detector via centre line regression and centre border probability

## Key ideas:

- taking advantage of the idea of TextMountain & FCOS. Predicting the center of a object is enough to separate texts.
- instance segmentation is not accurate and fast as regression methods , especial when we only use 1/4 feature map.

## Our Approach

- The FPN network outputs center-border probabilities and the top/bottom offsets at the same time.
- The regression targets are encoded as:

$$t^* = t - y * stride \quad (2)$$

$$b^* = b - y * stride \quad (3)$$

where  $t$  and  $b$  are top/bottom coordinates,  $y$  is vertical location at the feature map

- opencv is applied to cluster active pixels.

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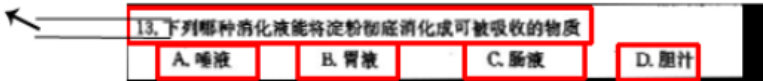
## Advantages:

- Fast: up to 10 fps (end to end tests, including read, write images and skew correction), 38 fps (only C<sup>2</sup>TD)
- Owing to the *link fuction* of centre lines, the methods can handle long text
- easy to handle small text lines

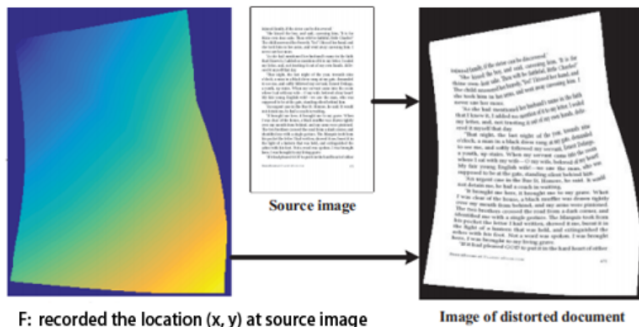


## C<sup>2</sup>TD: a fast and robust text detector via centre line regression and centre border probability

11 pixel



# Dewarp Text lines: Understanding Transform Grid

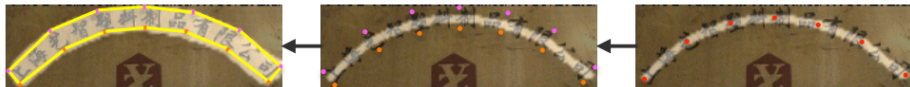


$$F^{-1} = 2 * referGrid - F \quad (4)$$

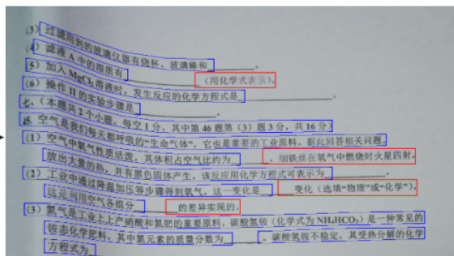
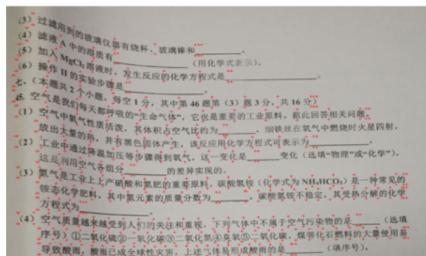
$F^{-1}$  recorded the locations at warped images.  
We use  $F^{-1}$  to match the text lines between source images and warped images.

# Detecting Text of Arbitrary Shapes

## Similar to LOMO

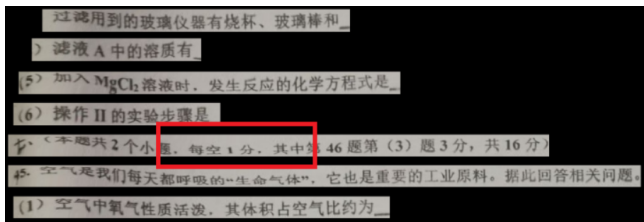
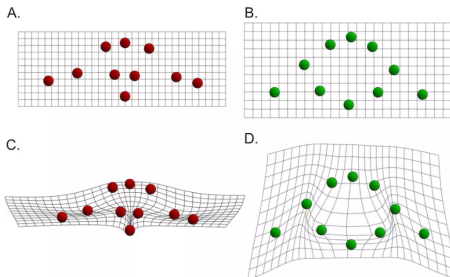


- Dynamic sampling: The sampling rate is related to the height of a text line.
- For a short textline, we treat it as a horizontal line.



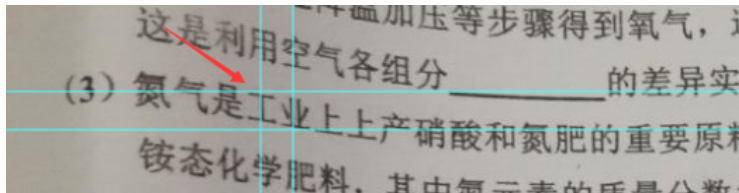
# Dewarp text lines based on key points

- Thin Plate Spline (TPS): Is TPS transformation too sensitive?

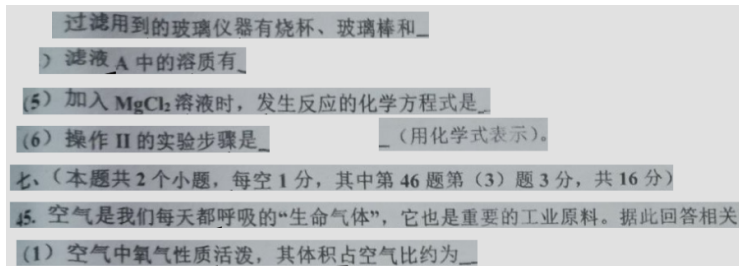


# Our solution: Piecewise Affine Transformation

- Parallel lines remain parallel after an affine transformation
- Does parallel lines remain parallel in real wrapped images?

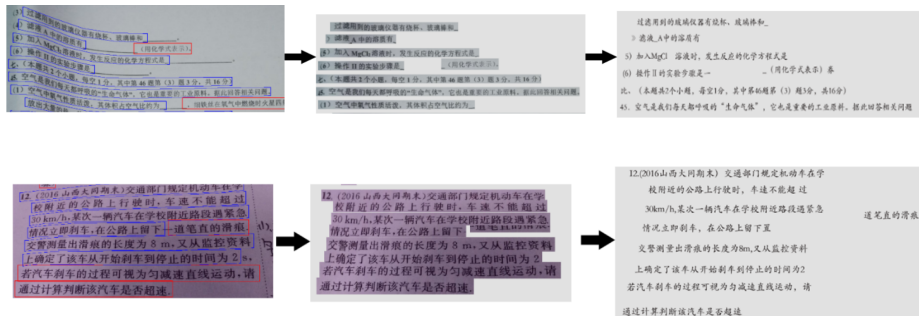


- Piecewise Affine transformation shows a better performance based on our observation.



# End to End Recognition

CNN based recognition system show a robust performance towards slightly distorted lines.



How to apply hard mining and weighting?