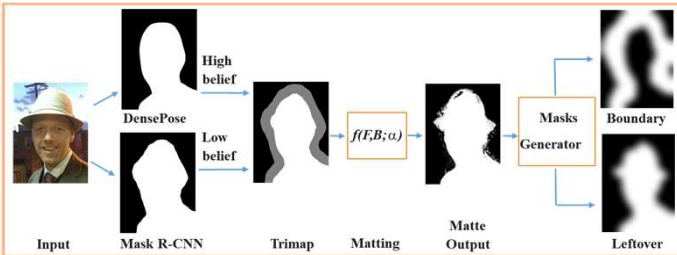


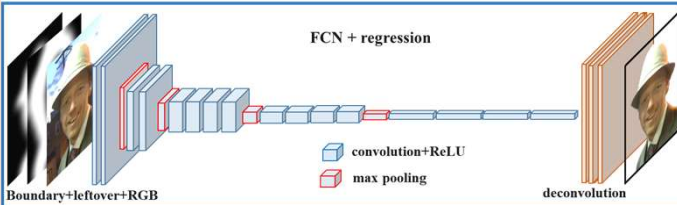
INTRODUCTION

- This paper introduces a novel framework for portrait segmentation with high precision on boundary area.
- Our proposal is based on:
 - A trimap generated by fusing information coming from Mask R-CNN and DensePose algorithms.
 - An alpha matting algorithm runs over the previous trimap
 - Portrait is refined by a Fully Convolutional Network (FCN) plus a transpose convolution.
- Tested on two evaluation datasets:
 - PFCN, considered the largest publicly available dataset
 - COCO dataset

OVERVIEW



Alpha matting based on Mask R-CNN and Dense pose



Matting refinement by FCN plus regression

FLASH TEXT

- Four stages are defined:**
 1. Trimap generation
 2. Alpha matting
 3. Boundary and leftover masks generation
 4. FCN plus regression

ATTENTION

Algorithm 1. Boundary mask generation

Input: binarized matte image I , kernel size $th1$
 $dilate_layer = dilate(I, th1)$
 $erode_layer = erode(I, th1)$
 $boundary_mask = dilate_layer - erode_layer$
 $boundary = GaussianBlur(boundary_mask, (2*th1)+1)$
Output: boundary

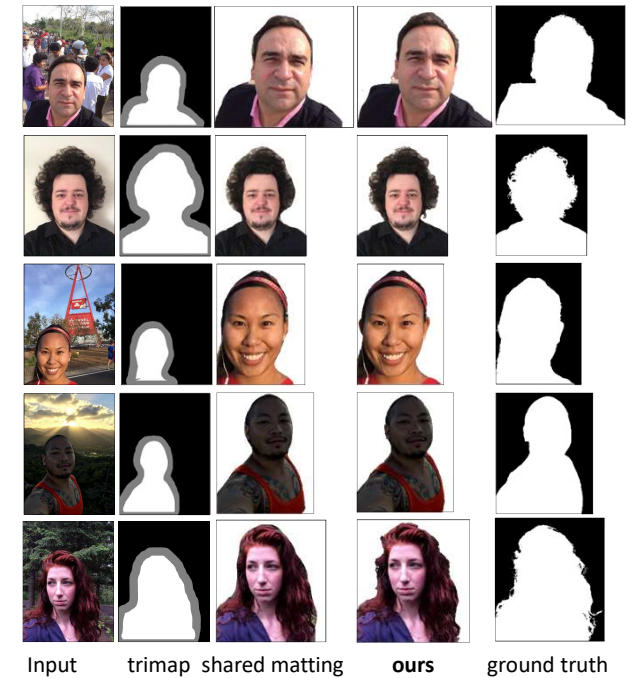
Algorithm 2. Leftover mask generation

Input: binarized matte image I , kernels size $th1$ and $th2$
 $Leftover_mask = erode(I, th2)$
 $leftover = GaussianBlur(leftover_mask, (2*th1)+1)$
Output: leftover

NUMERICAL RESULTS

Method	PFCN	COCO	COCO+
Shared Matting	94.0%	88.1%	94.0%
Portrait FCN+	95.9%	68.6%	
BSN	96.7%	77.7%	
ours	97.0%	89.6%	95.2%

RESULTS



CONCLUSION

- Improvement of matting techniques
- Hair segmentation is still a challenging issue
- Future work:** an end-to-end fully convolutional network taking as input the masks provided by Mask R-CNN and DensePose