

Unet for Semantic segmentation

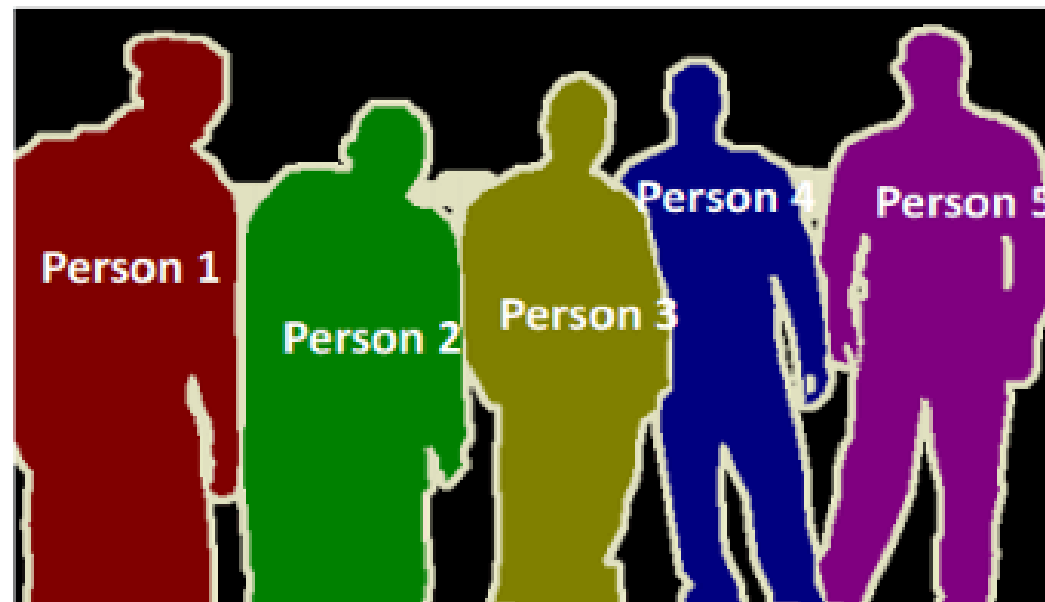
- . Architecture of Unet segmentation
- . Difference between upsampling and transpose convolution

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Difference between SS and IS



Semantic Segmentation



Instance Segmentation

Convolutional for downsampling

2	4	9	1	4
2	1	4	4	6
1	1	2	9	2
7	3	5	1	3
2	3	4	8	5

Image

X

1	2	3
-4	7	4
2	-5	1

Filter /
Kernel

=

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Feature

Transposed Convolutions

- Super Resolution -> Upscaling the input image to higher resolutions.
- Semantic Segmentation -> From RGB image input to class-based visualisation.



1. upsampling

- Repeats the rows and columns of the data by `size[0]` and `size[1]` respectively.
- No trainable parameters.

1. upsampling

1	2
3	4

1	2
1	2
3	4
3	4

1	1	2	2
1	1	2	2
3	3	4	4
3	3	4	4

1. upsampling

1	2
3	4

1	1	2	2
1	1	2	2
3	3	4	4
3	3	4	4

2. transpose convolution

- Convolution and upsampling.
- Trainable parameters.

2. transpose convolution

`Conv2DTranspose(1, (1, 1), strides = (2,2))`

No.filters

Filter size

stride

2. transpose convolution

Conv2DTranspose(1 , (1, 1) , strides = (2,2))

1	2
3	4

(w,b)

1x1 filter

$W = 2$

$b = 0$

2. transpose convolution

Conv2DTranspose(1 , (1, 1) , strides = (2,2))

1	2
3	4



1x1 filter
W = 2
b = 0

2. transpose convolution

Conv2DTranspose(1 , (1, 1) , strides = (2,2))

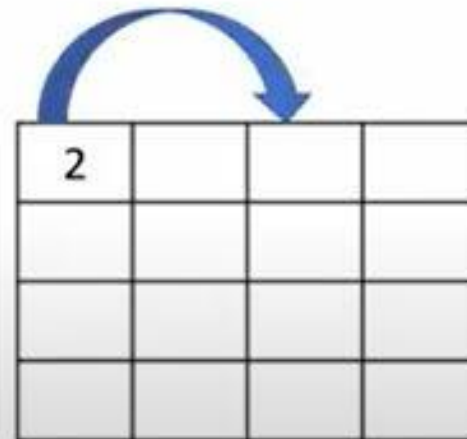
12	2
3	4

2			

2. transpose convolution

Conv2DTranspose(1 , (1, 1) , strides = (2,2))

1	2 ²
3	4



2. transpose convolution

Conv2DTranspose(1 , (1, 1) , strides = (2,2))

1	2
32	4



2	0	4	0

2. transpose convolution

Conv2DTranspose(1 , (1, 1) , strides = (2,2))

1	2
3	42

2	0	4	0
0	0	0	0
6	0	8	0
0	0	0	0

2. transpose convolution

Conv2DTranspose(1 , (1, 1) , strides = (1,1))

1	2
3	4

(w,b)

1x1 filter
 $W = 2$
 $b = 0$

2	4
6	8

transpose convolution with stride 1 = no upsampling

U-Net

