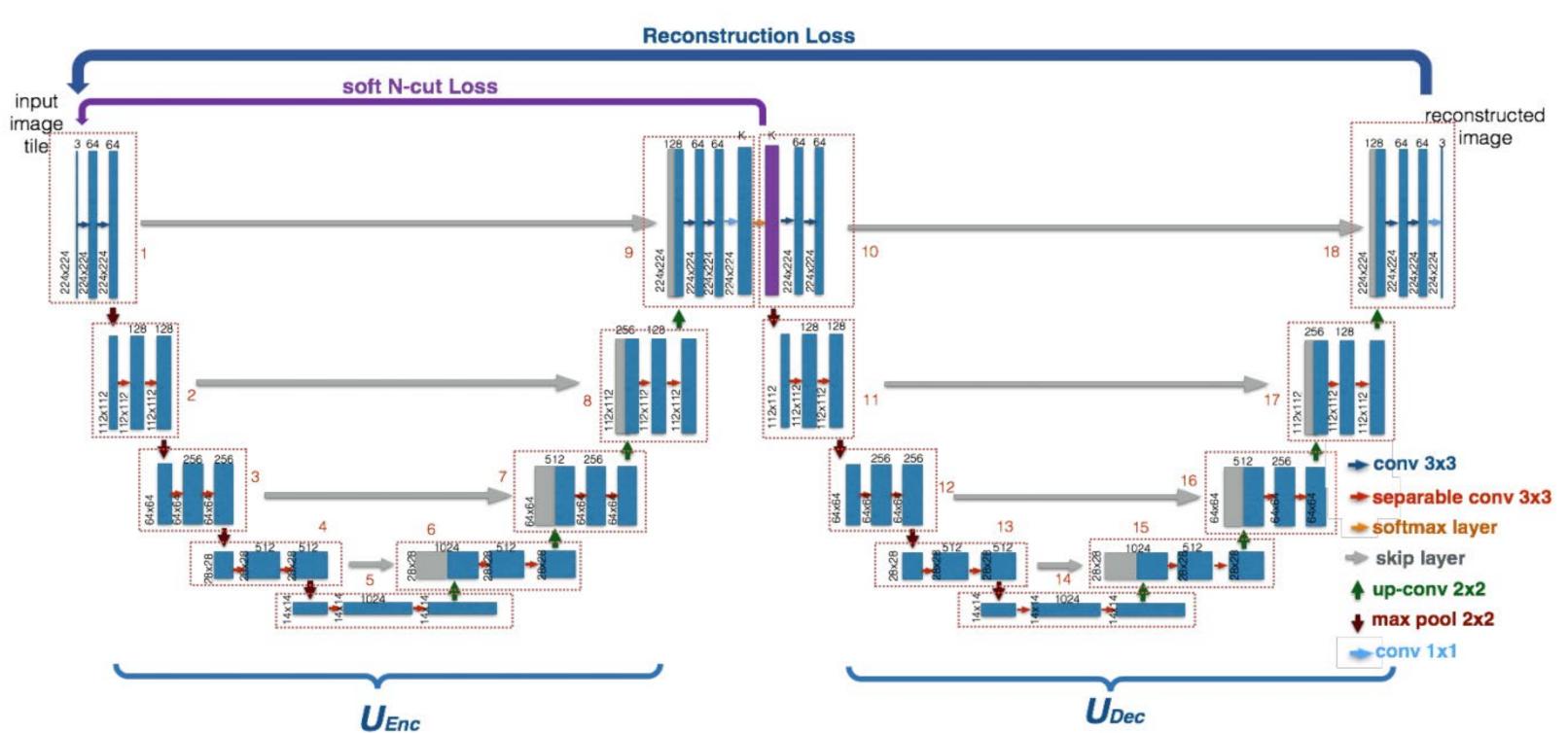
# W-Net: A Deep Model for Fully Unsupervised Image Segmentation Simon Tulling, Natalia Karpova

### General W-net architecture



W-Net is an auto-encoder consisting of two autoencoders. It consists of two connected auto-encoder networks, both having almost similar architecture. The only difference are the amount of channels in the input and output images. The first auto-encoder encodes the image into k segmented image, afterwards the second auto-encoder attempts to reconstruct it back into the original image using the segmented image.

Posprocessing is applied to an image after initial segmentation for the encoder is done. Postprocessing includes 2 main steps, namely:

> 1. Conditional Random Field 2. Hierarchical Segmentation

CRF is used to increase smoothness within a segmented image such that the final outcome will

have sharper boundaries but more even

reconstruction inside regions.

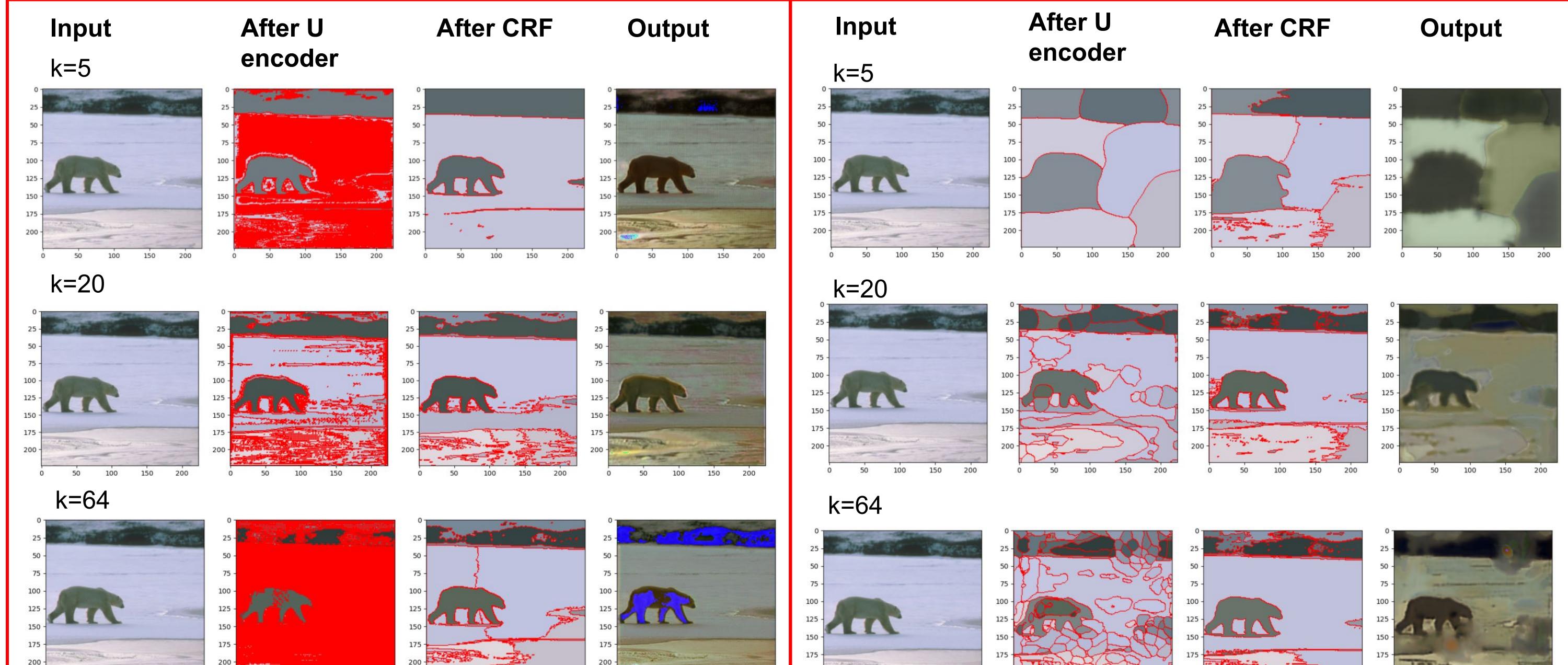
### Two types of losses

The paper makes use of 2 different loss functions. The main reconstruction loss is calculated in the end of W-Net pipeline:  $J_{reconstr} = \|\mathbf{X} - \mathbf{U}_{\mathbf{Dec}}(\mathbf{U}_{\mathbf{Enc}}(\mathbf{X}; W_{Enc}); W_{Dec})\|_2^2$ 

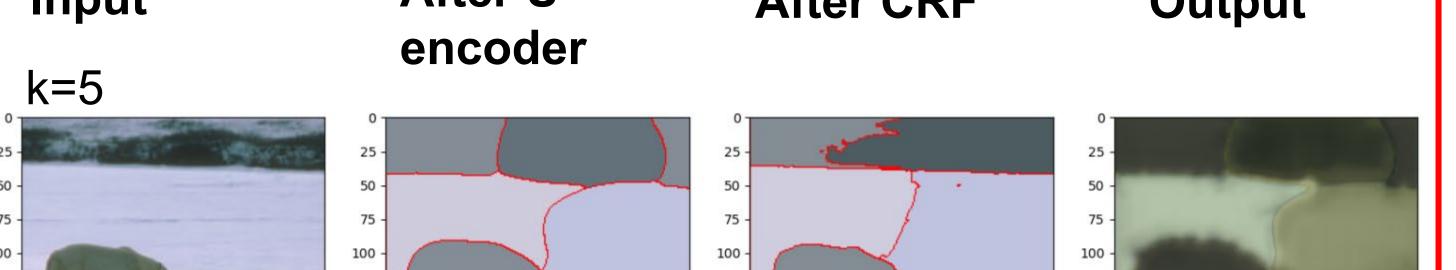
In addition to the reconstruction loss, soft N-Cut Loss is used on the output of encoder:

## Final results: white bear image example

Only reconstruction loss



$$J_{soft}(V,K) = \sum_{k=1}^{K} \frac{cut(A_k, V - A_k)}{assoc(A_k, V)}$$
$$= K - \sum_{k=1}^{K} \frac{assoc(A_k, A_k)}{assoc(A_k, V)}$$
$$= K - \sum_{k=1}^{K} \frac{\sum_{u \in V, v \in V} w(u, v) p(u = A_k) p(v = A_k)}{\sum_{u \in A_k, v \in V} w(u, v) p(u = A_k)}$$
Reconstruction and Soft N-Cut loss



#### **Conclusions:**

2.

3.

4.

- When using reconstruction loss only, image reconstruction is better, yet segmentation is not



#### Amount of classes around 20 seems to be the best k values

#### CRF does heavy lifting

We would like to thank our supervisor Attila Lengyel

for his help during this project

