

Single Image Super-Resolution Based on Capsule Neural Networks

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Computer Vision Tasks

Computer Vision Tasks



MRI of ankle¹



License plate²



Amazon deforestation³

¹Super Resolution Techniques for Medical Image Processing

²Beyond Human-level License Plate Super-resolution with Progressive Vehicle Search and Domain Priori GAN

³The Earth Observatory

Computer Vision Tasks

Computer Vision Tasks



LR MRI of ankle¹



LR license plate²

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Computer Vision Tasks



LR MRI of ankle¹



HR MRI of ankle¹



LR license plate²



HR license plate²

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"Enhance"

"Enhance"



Increasing resolution as seen in fiction¹

¹Adapted from [CSI Zoom Enhance on YouTube](#)

Upscaling

Upscaling



LR image¹

¹A database of human segmented natural images and its application to evaluating segmentation algorithms and measuring ecological statistics

Upscaling



LR image¹

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Upscaling



LR image¹



Upscaled image

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Super-Resolution

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LR image¹

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Super-Resolution



LR image¹

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Super-Resolution



LR image¹



Super-resolution image

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Super-Resolution

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- Most solutions use CNNs
 - SRCNN, EDSR, RDN, RCAN, WDSR, SRGAN, ESRGAN, ...
 - Good visual quality

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- Most solutions use CNNs
 - SRCNN, EDSR, RDN, RCAN, WDSR, SRGAN, ESRGAN, ...
 - Good visual quality
- CNNs have known drawbacks

Capsule Networks

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 - Inability to identify spatial hierarchy between elements
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- Proposes to solve some of the main flaws found in CNNs
 - Inability to identify spatial hierarchy between elements
 - Lack of rotation invariance
- Inspired by the human visual system
- Achieved good results in classification and segmentation tasks

Why "Capsule"?

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"Instead of aiming for viewpoint invariance in the activities of "neurons" that use a single scalar output to summarize the activities of a local pool of replicated feature detectors, artificial neural networks should use local "capsules" that perform some quite complicated internal computations on their inputs and then encapsulate the results of these computations into a small vector of highly informative outputs."¹

Convolution VS Capsule

Convolution VS Capsule

Capsules' Reconstructions

Properties	Reconstruction from capsules
Scale and thickness	6 6 6 6 6 6 6 6 6 6
Localized part	6 6 6 6 6 6 6 6 6 6
Stroke thickness	5 5 5 5 5 5 5 5 5 5
Localized skew	3 3 3 3 3 3 3 3 3 3
Width and translation	2 2 2 2 2 2 2 2 2 2
Localized part	4 4 4 4 4 4 4 4 4 4

Capsules Networks

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- Capsules originally were used in classification tasks
 - Achieved state-of-the-art results

Capsules Networks

- Capsules originally were used in classification tasks
 - Achieved state-of-the-art results
- Capsules networks have been explored in other tasks
 - Object detection
 - Image segmentation
 - Visual question answering

Capsules Networks

Capsules Networks

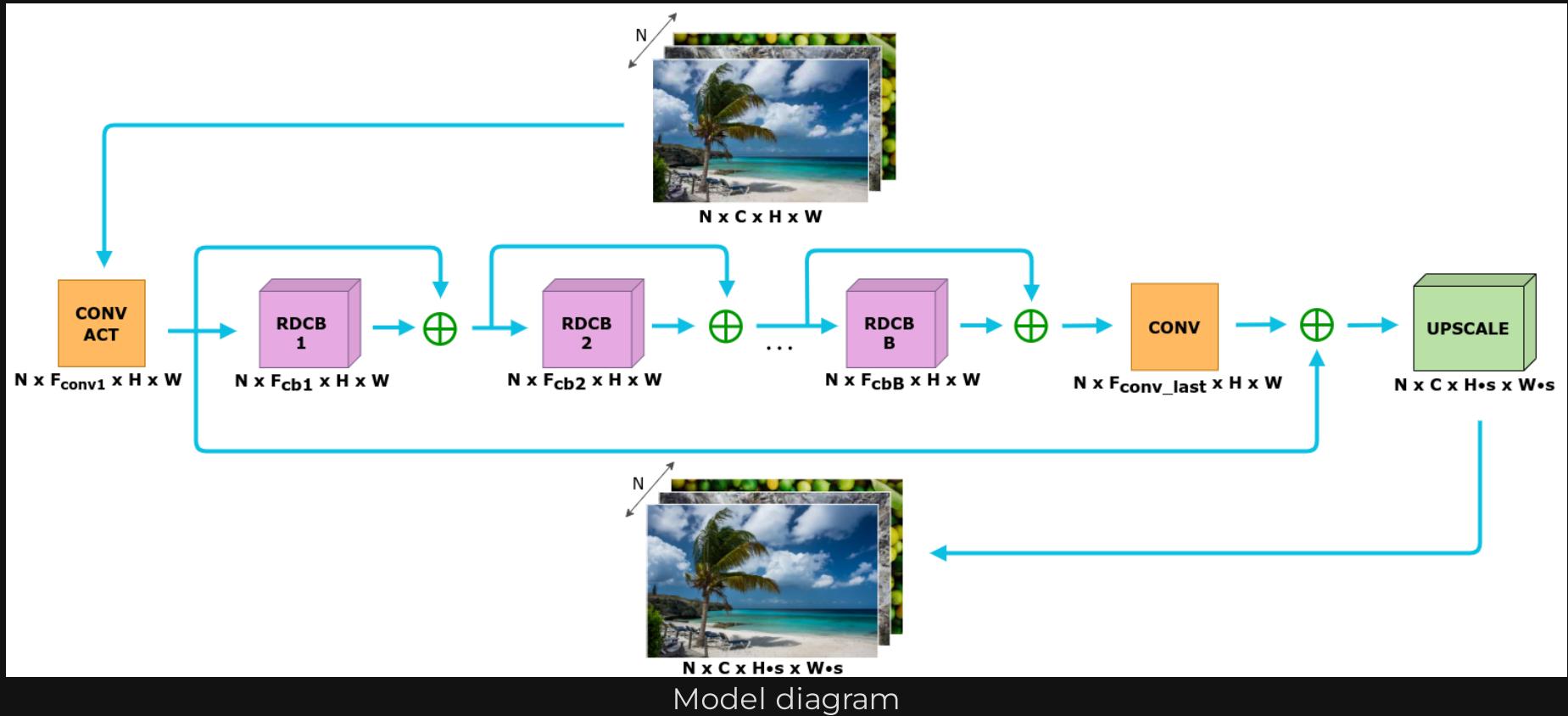
- Few explorations in SISR tasks
 - Little modifications to the original CapsNet

Capsules Networks

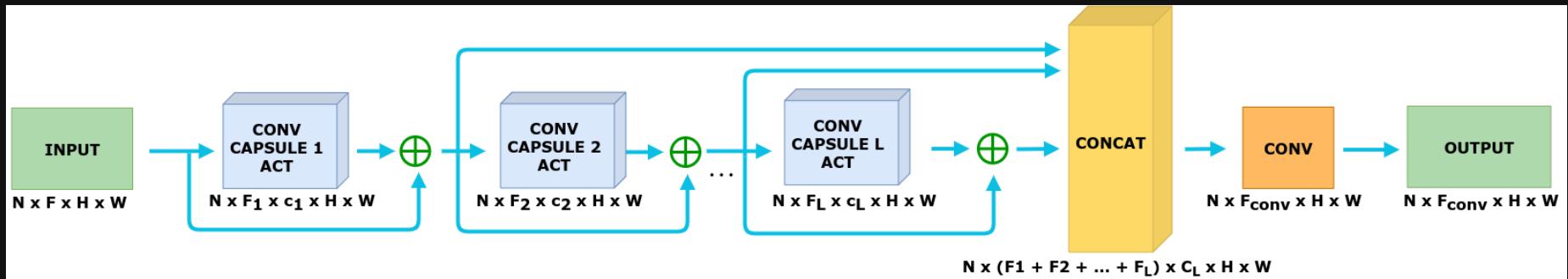
- Few explorations in SISR tasks
 - Little modifications to the original CapsNet
- Novel concepts have been applied to CapsNets
 - Different capsules types
 - New routing algorithms

SRCaps

SRCaps

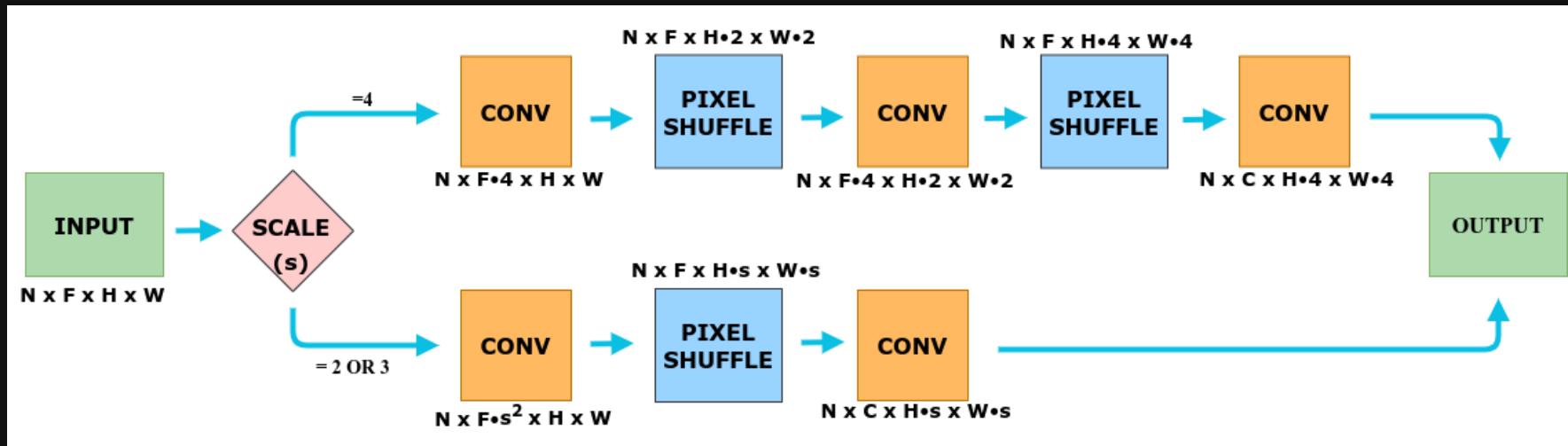


SRCaps



Capsblock diagram

SRCaps



UPNet diagram

Experiments Setup

Experiments Setup

- Training
 - DIV2K training set
 - Losses: L_1 , SSIM, MS-SSIM, L_1 after a few RDCBs, L_1 + edge map, 3-PSNR, 3-SSIM, adaptive loss
 - Model configuration: refer to the paper

Experiments Setup

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 - Model configuration: refer to the paper
- Validation
 - DIV2K validation set, Set5, Set14, BSD100 (B100), Urban100
 - Metrics: PSNR, SSIM, MS-SSIM, FLIP

Adaptive Loss

α Loss function

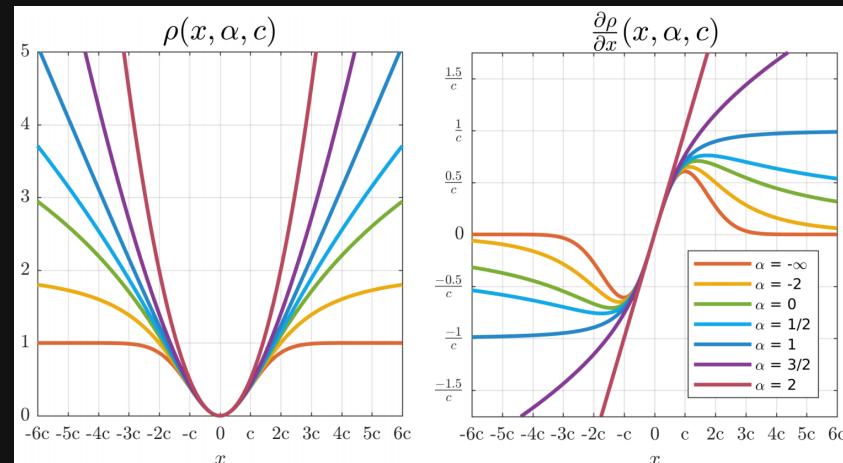
2 L2

1 Charbonnier / pseudo-Huber / L1-L2

0 Cauchy/Lorentzian

-2 Geman-McClure

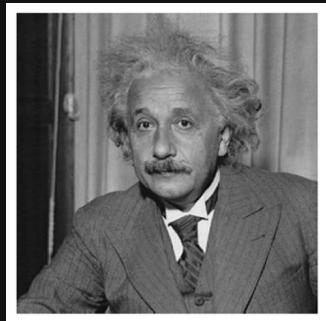
$-\infty$ Welsch/Leclerc



The general loss function (left) and its gradient (right) for different values of its shape parameter α ¹

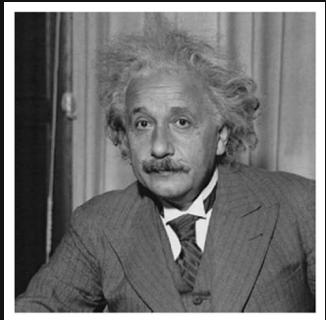
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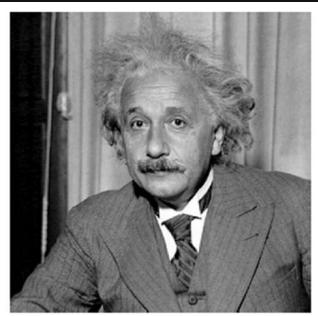


Reference

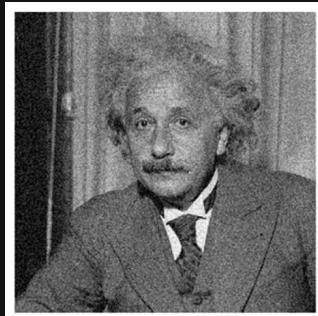
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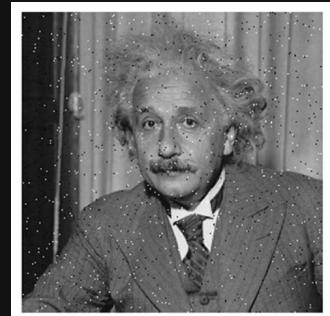
Reference



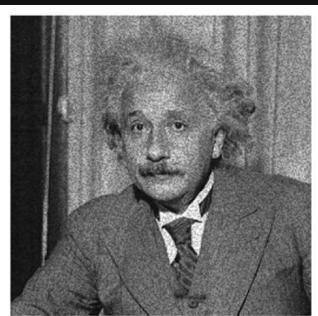
Contrast
enhanced



Gaussian noise



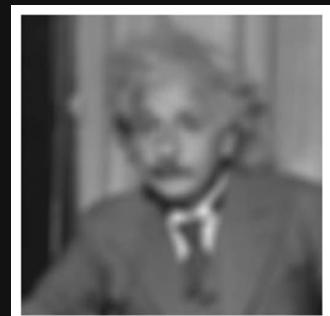
Salt-pepper noise



Speckle noise



JPEG



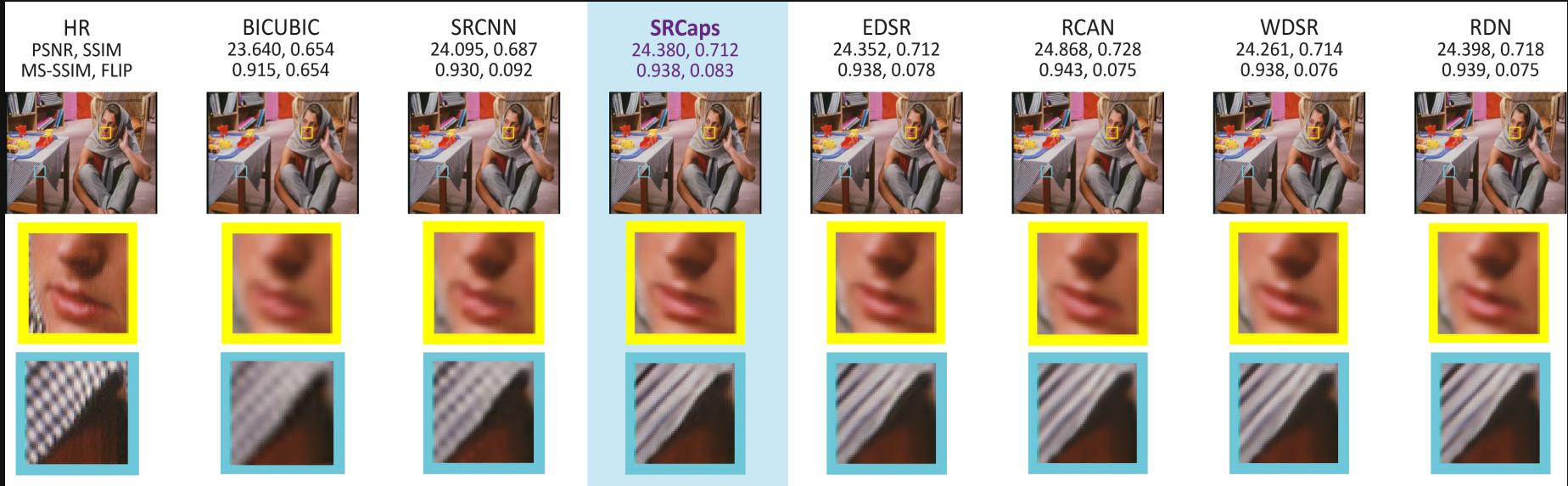
Blurred

Model Comparison

	SRCaps	EDSR	RCAN	WDSR	RDN	SRCNN
Number of Parameters	15M	1.5M	12.6M	4.8M	22.3M	20.1K
Number of Blocks	7	16	10 × 16	16	16	1 (not residual)
Number of Layers per Block	4	2	3	3	8	3
Dense Connections	✓	✗	✗	✗	✓	✗
Uses mean RGB	✗	✓	✗	✓	✗	✗
Sub-pixel Convolution	✓	✓	✓	✓	✓	✗
Loss Function	Adaptive	L1	L1	L1	L1	L1

Results

Results



Model results for “barbara” image from Set14 dataset

Results

HR PSNR, SSIM MS-SSIM, FLIP	BICUBIC 23.903, 0.706 0.935, 0.118	SRCNN 24.568, 0.736 0.947, 0.107	SRCaps 25.980, 0.792 0.963, 0.088	EDSR 26.255, 0.805 0.966, 0.075	RCAN 26.955, 0.828 0.970, 0.075	WDSR 26.692, 0.821 0.969, 0.070	RDN 26.826, 0.825 0.970, 0.070
							
							
							

Model results for “0891” image from DIV2K dataset

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 - nonlinearity function applied may be a limiting factor

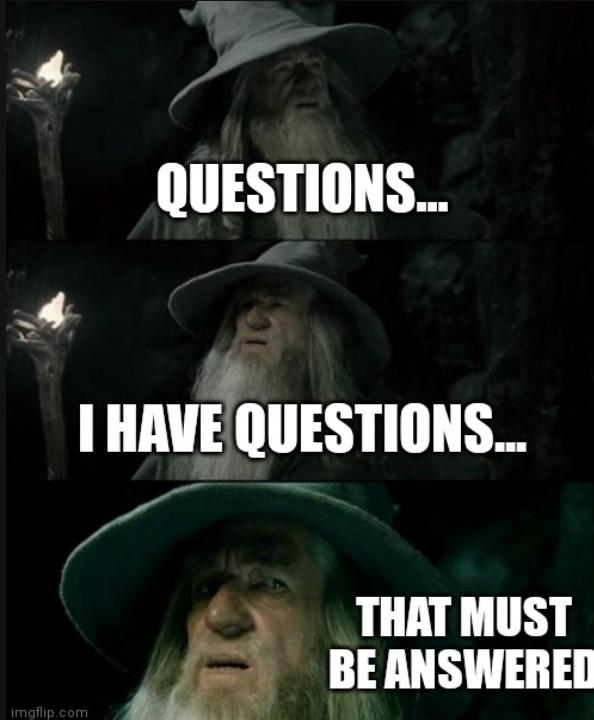
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- Highlights
 - RCAN ability to create smooth edges
 - adaptive loss function
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- Highlights
 - RCAN ability to create smooth edges
 - adaptive loss function
 - metrics not exactly reflect visual quality
- Future research
 - replace the composition of the UPNet
 - new non-linearity and routing functions for the capsules
 - novel capsule models

Questions?



Gandalf has questions¹