

# DarioNet

Improve Tiny Images Classification Through  
Super Resolution (and vice versa)

*The DarioNet team* 🏰🍳🐟:

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# DATA SOURCE



## Train

15% of ImageNet  
**Training set**  
(~195K images  
~500x400, ~21Gb)

## Validation

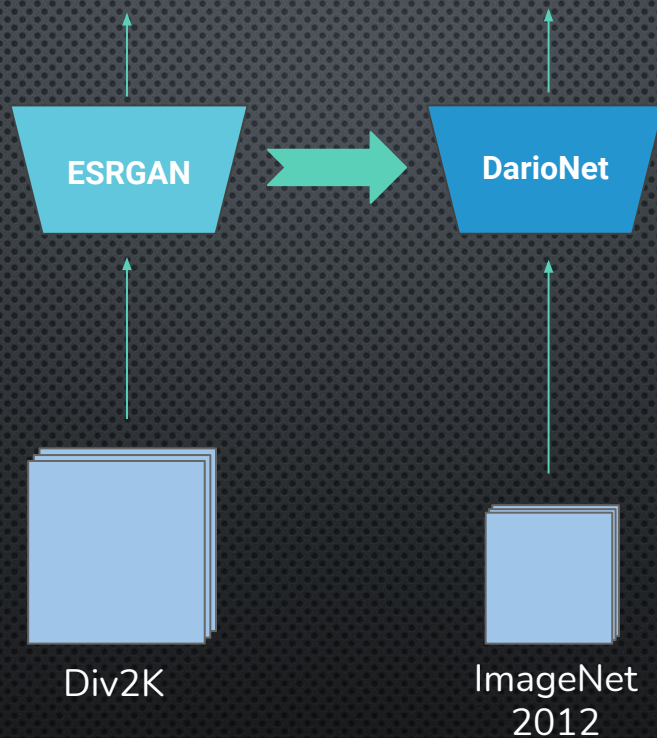
50% of ImageNet  
**Validation set**  
(~25K images ~500x400,  
~3.5Gb)

## Test

50% of ImageNet  
**Validation set**  
(~25K images ~500x400,  
~3.5Gb)

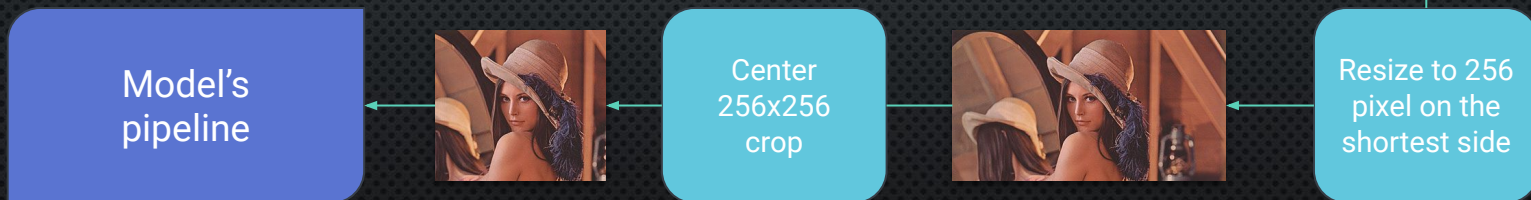


# ESRGAN vs DarioNet



# Images preprocessing

- **Data augmentation** seems to worsen the results and is not so needed since we have a very large dataset of images
- Previously we had random horizontal flip, random vertical flip and random crop at the end





# End-to-end models pipelines

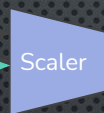
**Model 1** (no downscaling/  
super resolution)



ResNet50\*



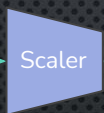
**Model 2** (bilinear  
upscaling)



ResNet50\*



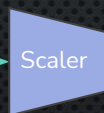
**Model 3** (ESRGAN  
super resolution)



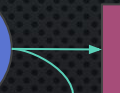
ResNet50\*



**Model 4** (DarioNet  
super resolution)



ResNet50\*



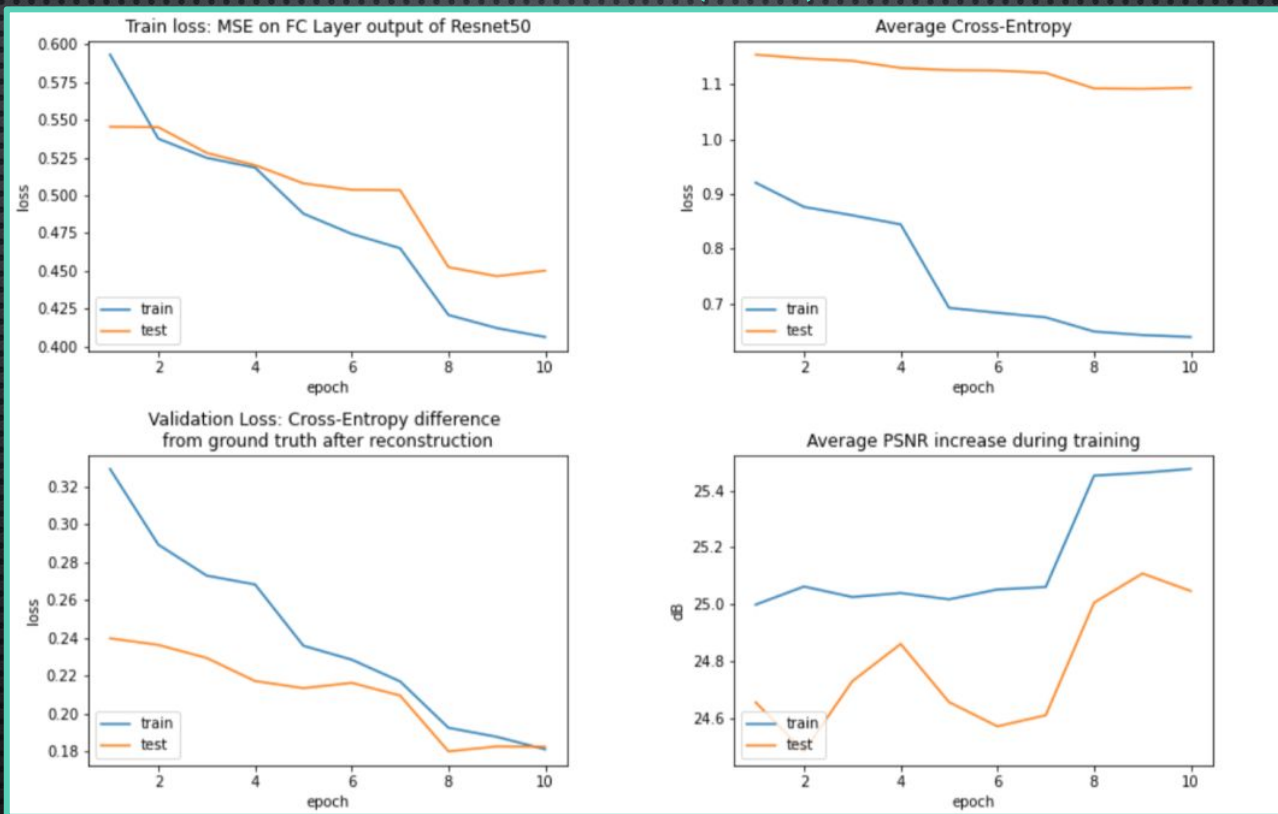
classification loss

# Losses (1/2)

- **PSNR** ✗
  - indeed one of the most used loss for this application
  - considered unstable for training and seems to lead to blurred images
- **L1** ✗
  - Manhattan distance
  - lead to blurred images
- **Distillation with scores** ✓
  - in our case, MSE between ResNet's scores
  - focussing on the classification
  - highest scores
- **Combination of various losses** ✗
  - tradeoff between aesthetics and performances
  - problem: different losses have different order of magnitude



## Losses (2/2)



## Results (1/4)

Is it possible to play just right now with DarioNet:

- Apply Super Resolution to an image via [Google Colab](#)
- Take a look at test.py and try\_darionet.py in our [GitHub repo](#)

**Model 1** (no downscaling/  
super resolution)



**Model 2** (bilinear super  
resolution)



**Model 3** (ESRGAN super  
resolution)

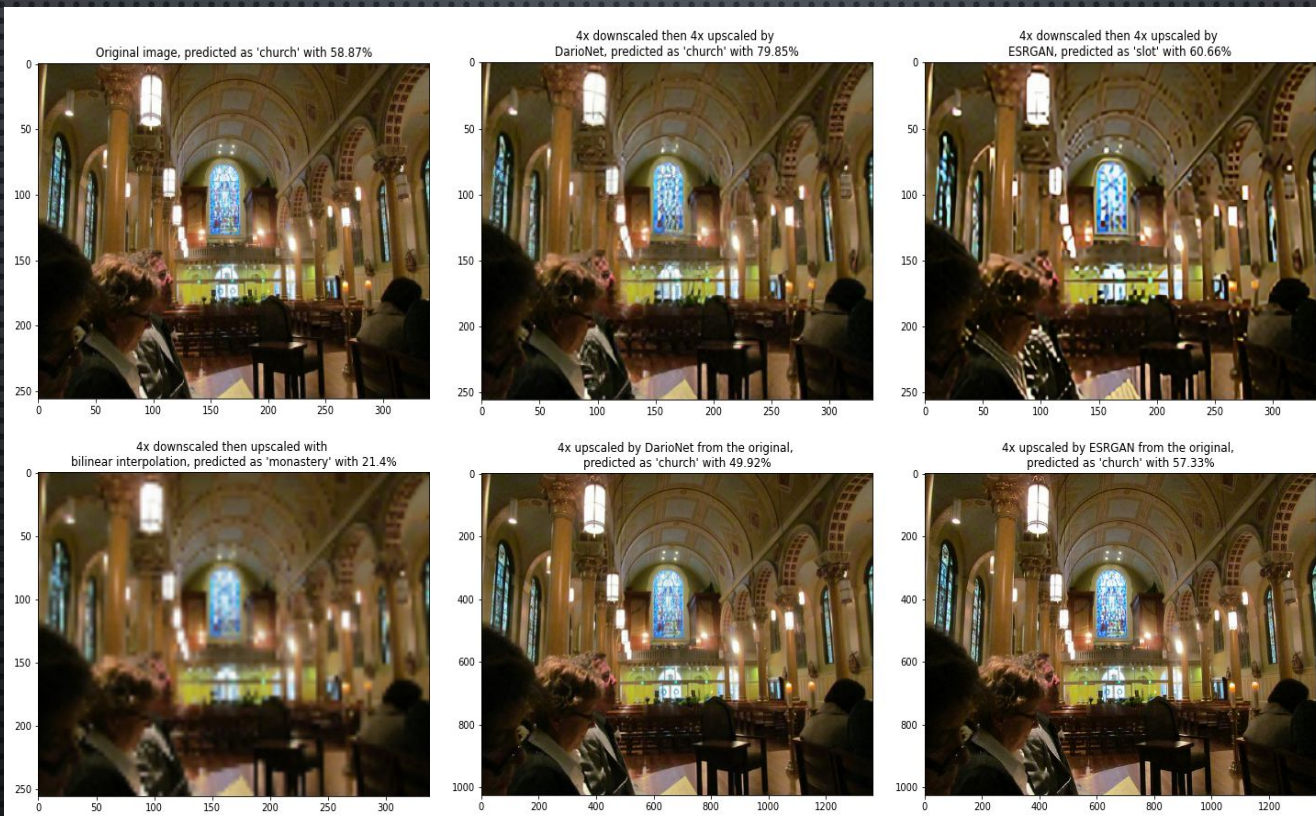


**Model 4** (DarioNet super  
resolution)





## Results (2/4)



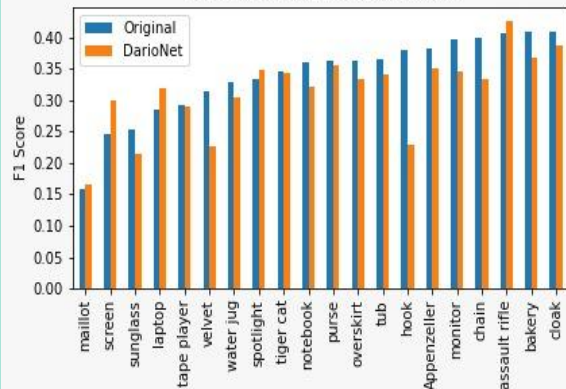
## Results (3/4)

Name of the model	Average cross entropy loss	Average PSNR (dB)	Accuracy	Total time on test (minutes)
<b>Model 1</b> (Original images)	0.9112	inf	0.7676	1
<b>Model 2</b> (bilinear upscaling)	1.9864	25.2466	0.5446	1
<b>Model 3</b> (ESRGAN super resolution)	1.6946	22.5088	0.6063	14
<b>Model 4</b> (DarioNet super resolution)	1.0931	25.2211	0.7275	14

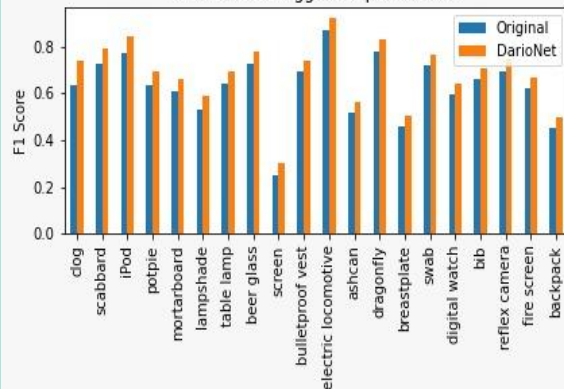


# Results (4/4)

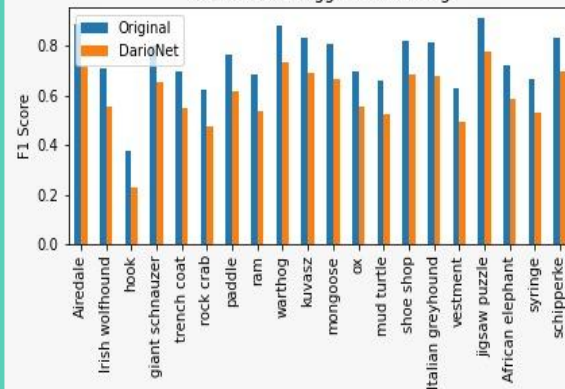
Performances on Hardest Classes



Classes with Biggest Improvements



Classes with Biggest Worsening



# Conclusions and Future work

- **Starting idea:**  
evaluate the performances of super-resolution on different task.
- **Primary interest of our work:**  
improve the results on the classification task starting from low resolution inputs
- A step towards **image compression**
- **Train also ResNet50** (testing on tiny ImageNet dataset)
- train DarioNet on **new tasks** like semantic segmentation or object detection
- See the results when classification has **less classes to predict**



Thank you for the attention!!  
Questions?

- THE DARIONET TEAM 🏰🍷🐟