

# Super Resolution

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## Network Architecture

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I build 3-layered network for Super Resolution tasks(2x), 3x3 conv with ReLU as activation. (see **SuperResolutionNetwork** at `lib/network.py`)

Detail hyper parameter setup as below:

- Use Adam optimizer with `lr=.001` and `lr_decay=.0`.
- Use **Mean Squared Error** as loss
- Use PSNR as evaluation metric
- Initialize kernel(*random uniform*) and bias(*zeros*) in convolution layers.

## Dataset

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Use all `91` and `291` dataset. Create target image, random crop from image after random scale down and create source image using scale down and up by half.

## Evaluation

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I use PSNR as evaluation metric, (see **SuperResolutionNetwork.metric** at `lib/network.py`) using TensorFlow implementation, `tf.image.psnr`.

Implement **Custom Callback** (see **CustomCallback** at `utils/callbacks.py`) for logging loss, accuracy and sample images. For each interval, write inferenced image summary for one train set and all test set.

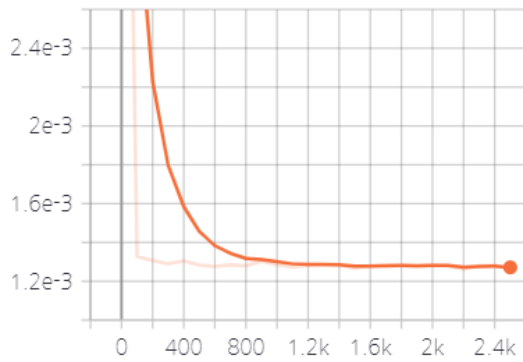
## Results

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See `./assets` directory for results. Train, prediction and test image is merged for easy compare (re-scale image, prediction image, ground truth image).

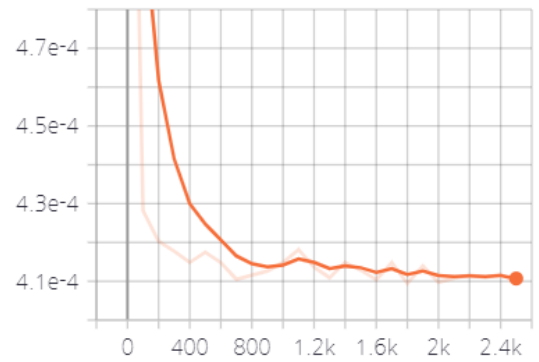
loss

loss



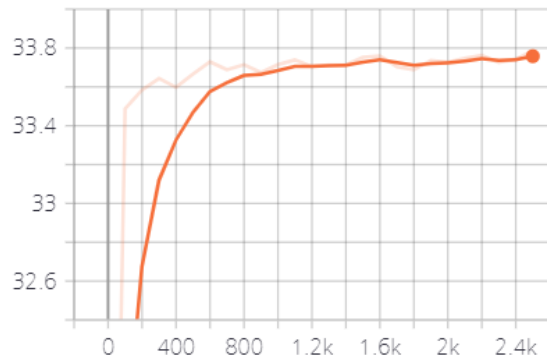
val\_loss

val\_loss



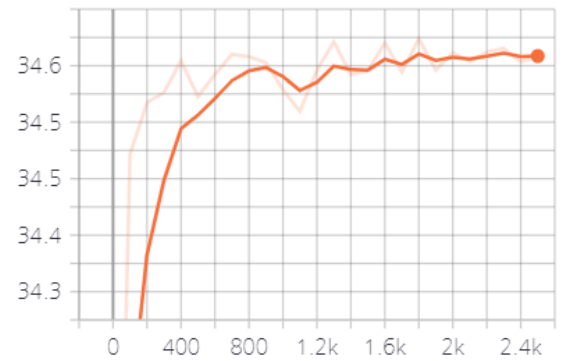
metric

metric



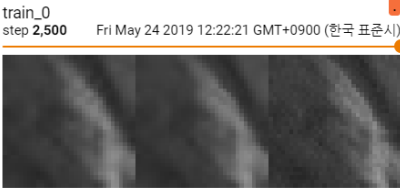
val\_metric

val\_metric



This is some train samples for check train is running well.

train\_0



train\_10



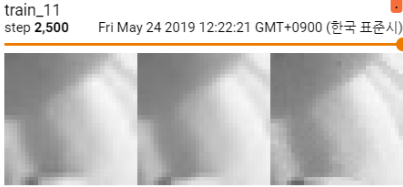
train\_72



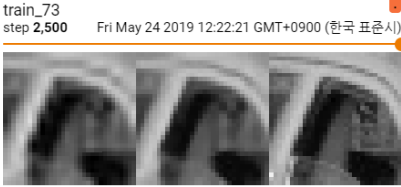
train\_1



train\_11



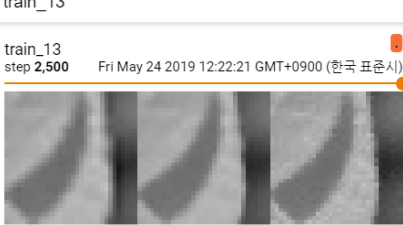
train\_73



train\_2



train\_12



train\_74

