



# Denoising radio traces with autoencoder on Tunka-Rex experiment

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#### Motivation



- Real background is not absolutely white
- It is hard to manually categorize each possible RFI
- The true pulse shape is known with low accuracy



#### Architecture





# Learning strategy

# Karbruhe Institute of Technology

#### Dataset

- 650k samples of Tunka background recorded in 2014-2017
- CoREAS simulations of Tunka-Rex signals
- Pulse is randomly located inside signal window (200 ns)
- Using single polarization
- Upsampling 16, using 4096 bins per trace (1280 μs)

#### Three different networks

- High amplitudes  $A > 200 \ \mu\text{V/m}$
- Medium amplitudes  $100 < A < 200 \ \mu\text{V/m}$
- Low amplitudes  $A < 100 \ \mu\text{V/m}$  present work
  - 11k samples + 1k augmented samples (signal is out of signal window)
  - 90% for training
  - 10% for control

# Signal normalization





### Issue with normalization



- Absolute amplitude of denoised trace is not reconstructed
- Re-normalization to scale of input trace can fix the problem



# Smoothing of the signal



- Reconstruction adds high-frequency numerical noise
- Savitzky-Golay filter is used for smoothing



# **Bandpass filtering**





#### Example of denoising





#### Few more examples







# Improving signal-to-noise ratio





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### Summary

#### **Present status**

- Autoencoder shows promising results
- The overall SNR is increased on the chosen dataset
- Prospects of investigation of spectral features

#### Next steps

- Fixing the issue with normalization
- More sophisticated criteria for estimation of performance is required
- Comparison of performance with matched filtering
- Investigation the different configurations: upsampling, trace length, size of dataset
- Including second polarization in analysis
- Alternative training in Fourier domain



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