

Learning-Based Shadow Mitigation For Terahertz Multi-Layer Imaging

P. Wang, T. Koike-Akino, A. Bose, Rui Ma, P. V. Orlik

Mitsubishi Electric Research Laboratories (MERL), Cambridge, MA, USA

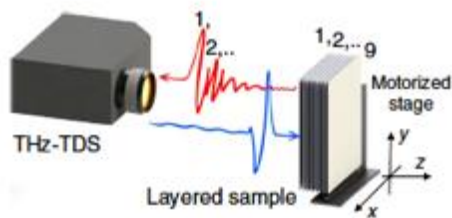
W. Tsujita, K. Sadamoto, H. Tsutada

Mitsubishi Electric Corporation Advanced Technology R&D Center, Amagasaki City, Japan

M. Soltanian

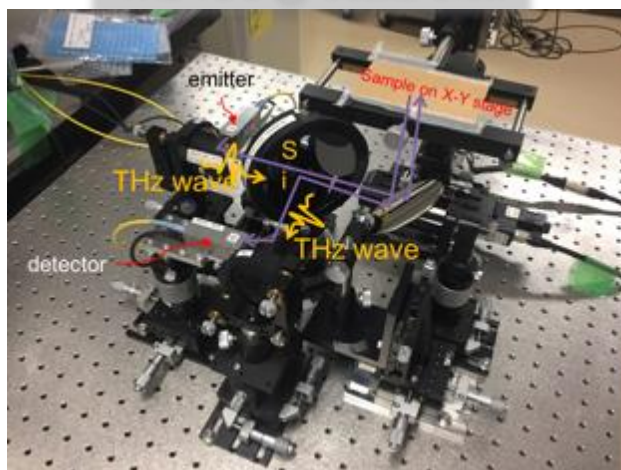
University of Illinois at Chicago, Chicago, USA

Raster Scanning Mode

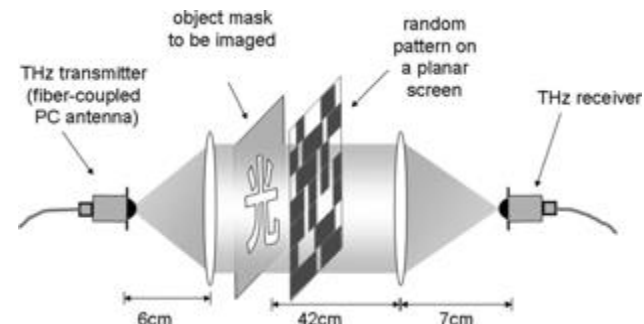


- Normal incident THz beam
- Mechanically move THz transceiver or samples
- Spot size: 1-mm radius at 1 THz
- Pulse width: 1-5 ps
- Sampling frequency: up to 5 THz
- Scanning rate: 100 Hz to 1000 Hz

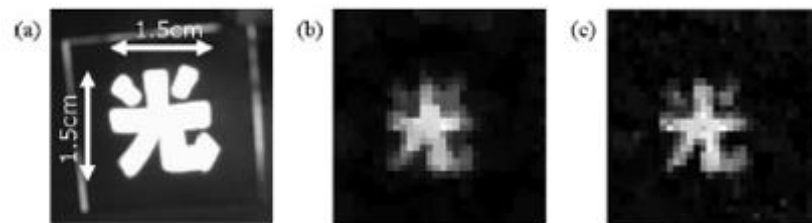
THz-TDS @ Osaka Univ.

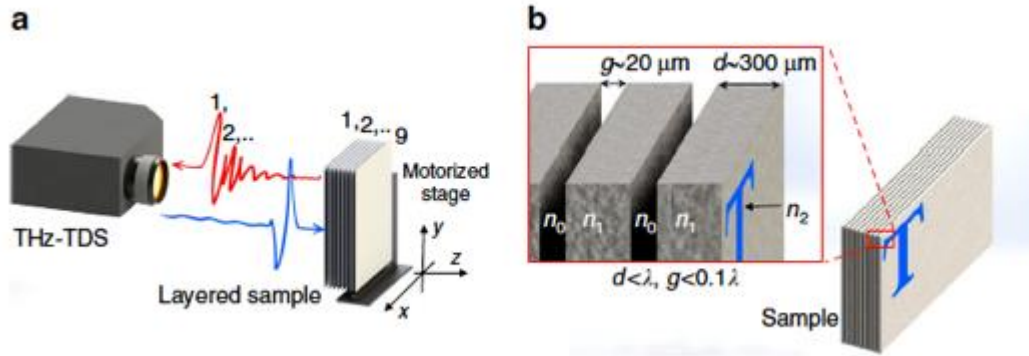


Compressed Scanning Mode

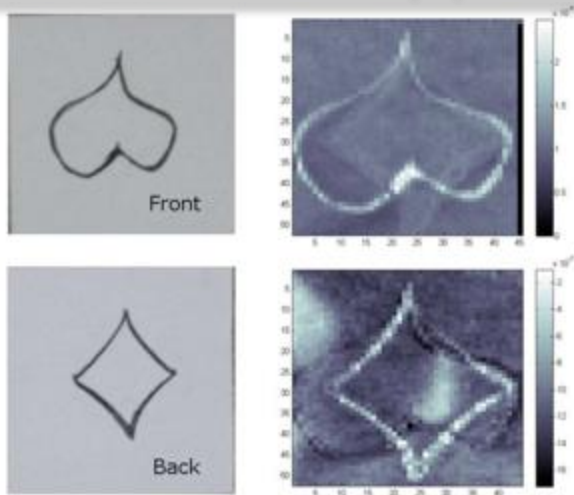


- Collimated THz beam
- Spatial light modulator (SLM) at THz band
- Focusing lens before THz detector
- No mechanical scanning
- Need random masks



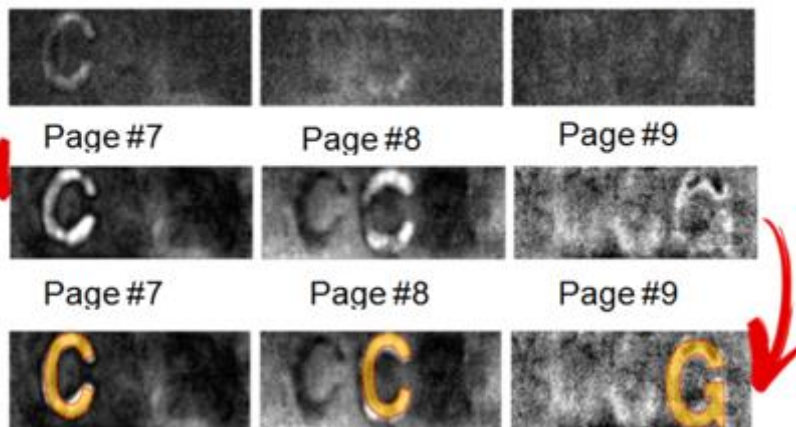


Ref. [1]: THz images of the front and back of the first sheet of paper



Ref. [3]: MIT THz See-Through Book

Measured time-domain E-field amplitude on page 7-9



Ref. [2]: THz image (c) of the painting *La oracion en el huerto*

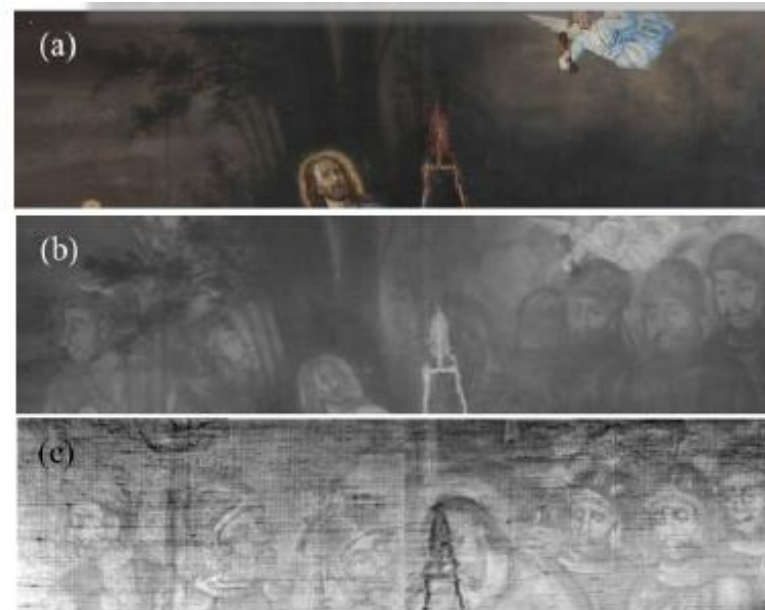


Fig. 3. a. Second detail of the painting *La oración en el huerto*. b. the infrared image of the detail. c. the THz record of the detail

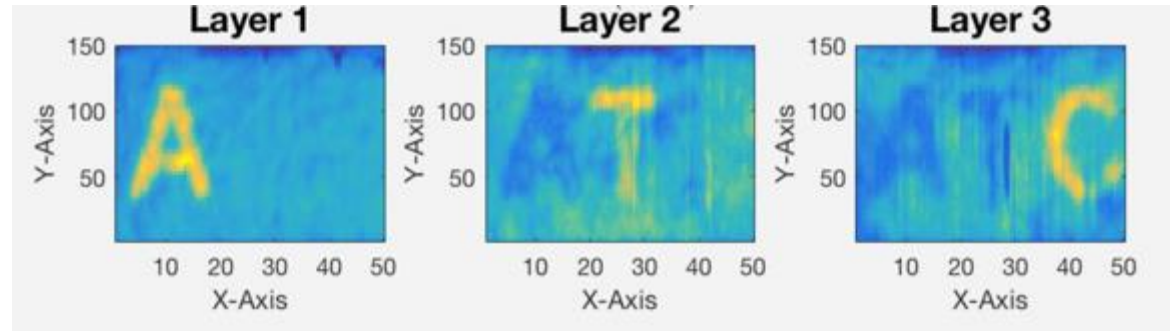
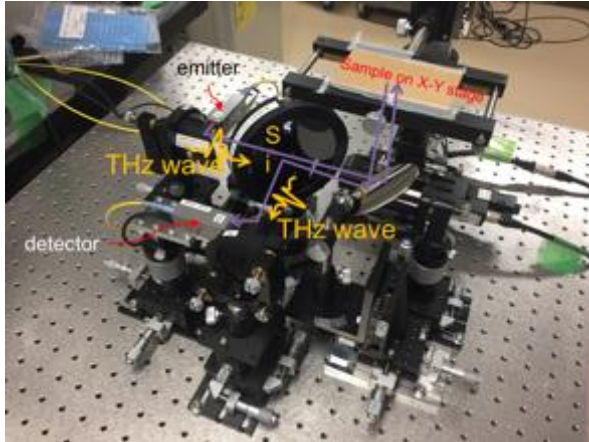
[1] G. C. Walker, et al., "Terahertz deconvolution," *Optics Express*, vol. 20, no. 25, pp. 27230–27241, Dec. 2012.

[2] C. L. K. Dandolo et al., "Contribution of terahertz time-domain analysis to art history: The case of the paintings of the Santo Entierro de Nuestro Señor Jesucristo altarpiece," 42nd IRMMW-THz, Cancun, Mexico, 2017

[3] A. Redo-Sanchez, et al., "Terahertz time-gated spectral imaging for content extraction through layered structures," *Nature Communications*, vol. 7, pp. 1–7, Sept. 2016.

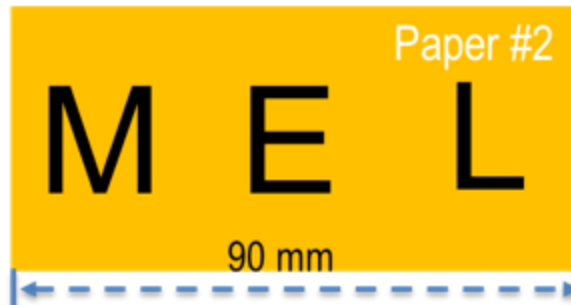
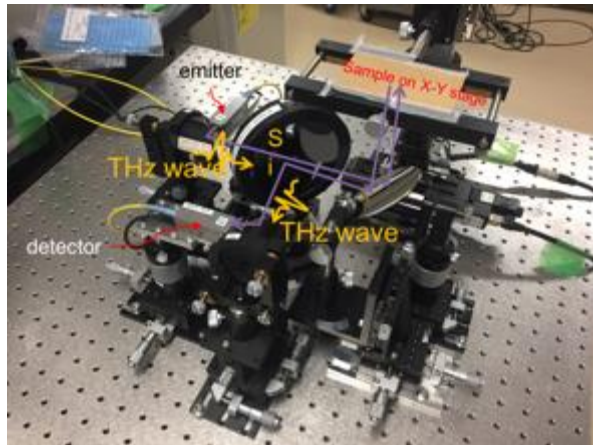
Shadow Effect (II)

Our own experiment at Osaka University



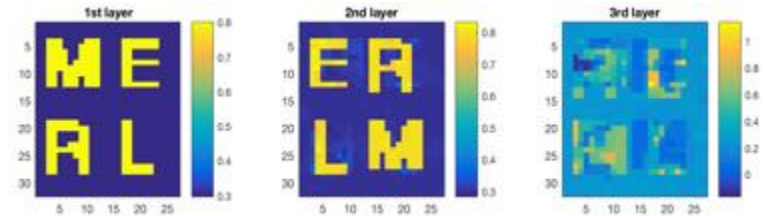
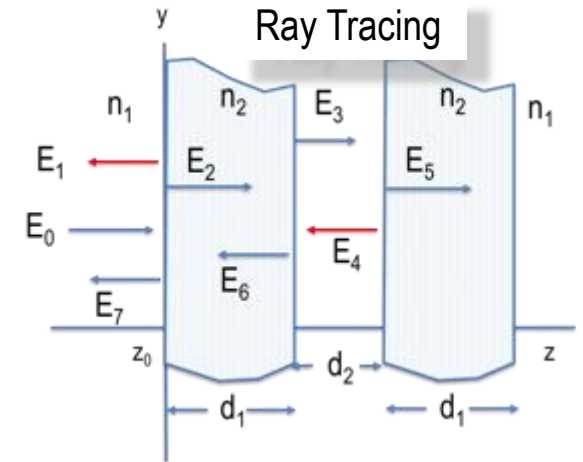
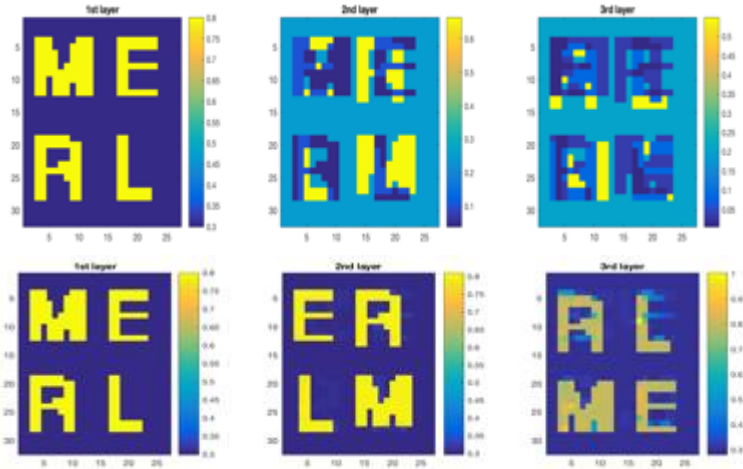
Shadow Effect (III)

Our own experiment at Osaka University



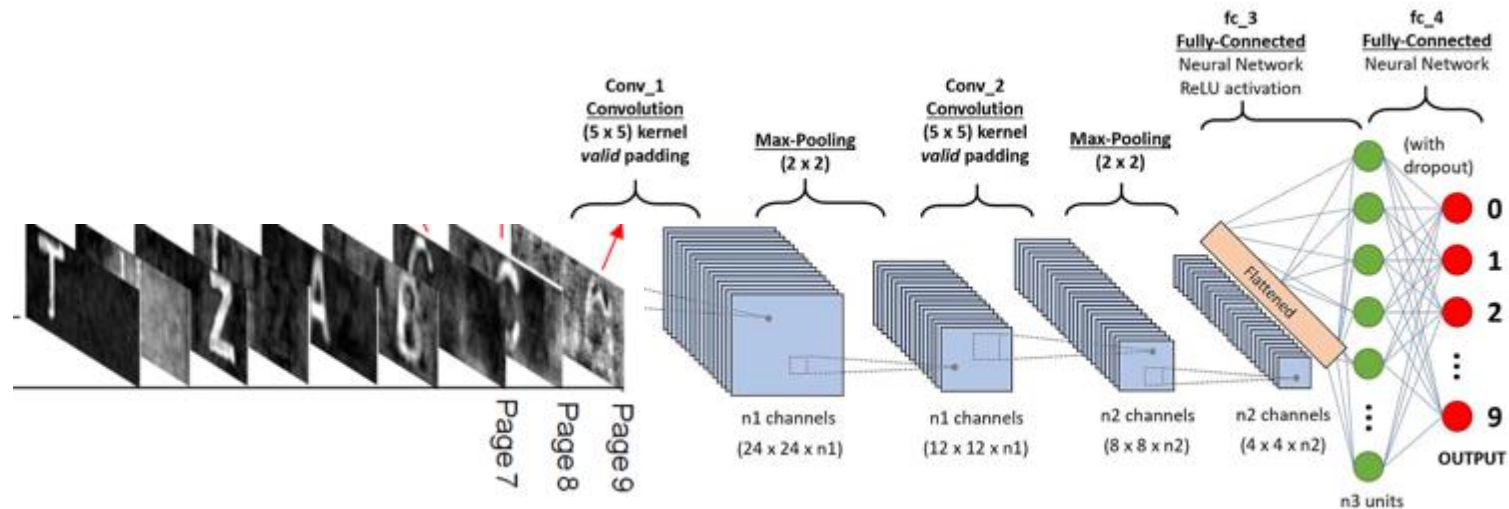
Ray-tracing model

$$x(t) = \rho_1 s \left(t - 2 \frac{z_0}{c} \right) + \rho_2 (1 - \rho_1^2)^2 s \left(t - 2 \frac{z_0}{c} - 2 \frac{n_2 d_1}{c} - 2 \frac{d_2}{c} \right)$$



Nonlinear scattering model

Image-domain learning approaches

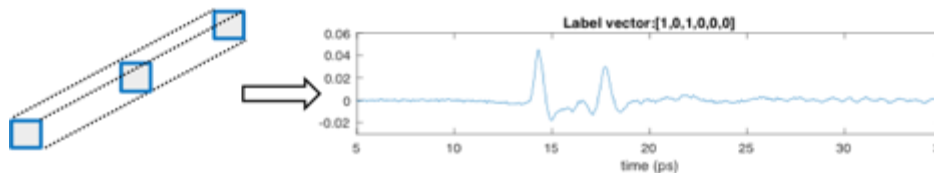
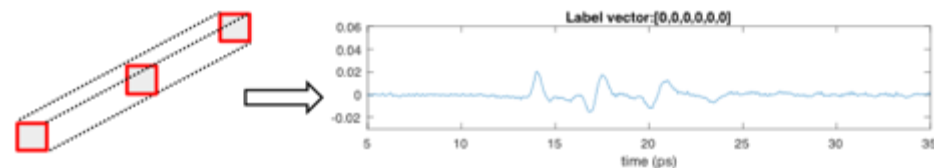
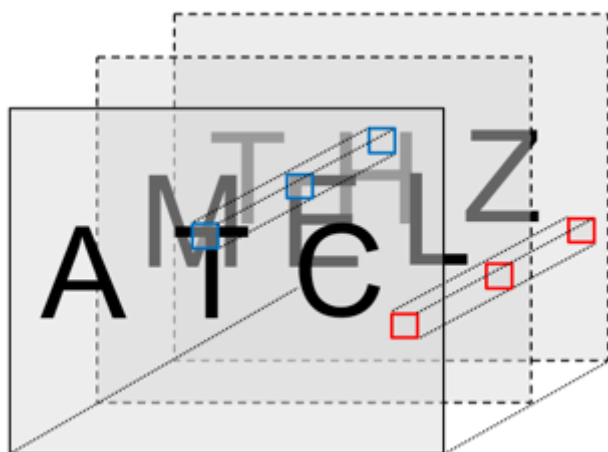


- Inputs: multi-layer THz images with shadow effects
- Labels: ground truth
- Convolutional neural networks (CNN) with multi-dimensional local convolution
- May work well with sufficient multi-layer THz images
- Challenges: multi-layer THz images are limited.

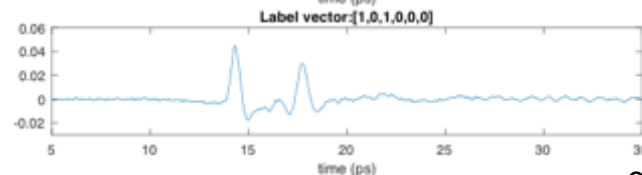
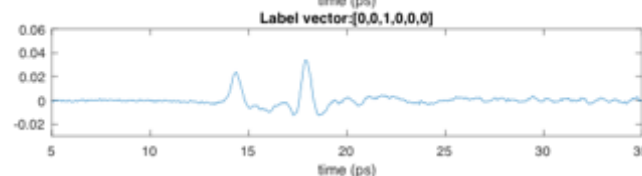
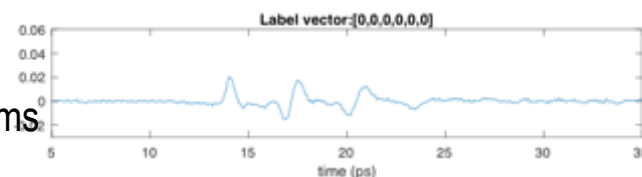
Voxel-domain learning approaches

Voxel-domain learning approaches

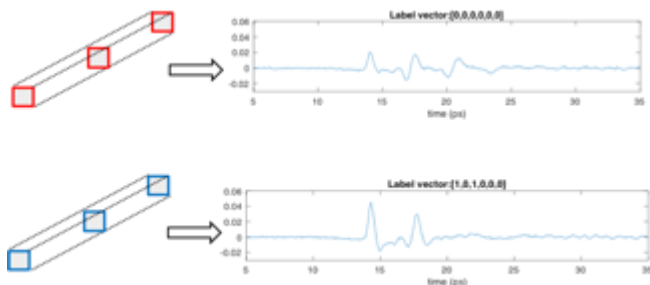
- Decompose multi-layer images into voxels
- Voxels are labeled by binary vectors (0/1): 0 denotes no content while 1 means there is a content
- The dimension of binary vectors is $2L$, where N is the number of layers



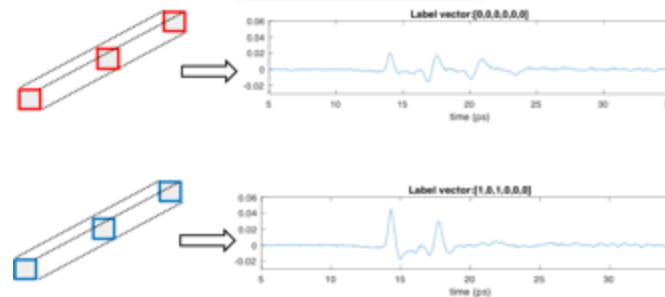
- Each voxel is attached to corresponding THz-TDS time-domain waveforms.



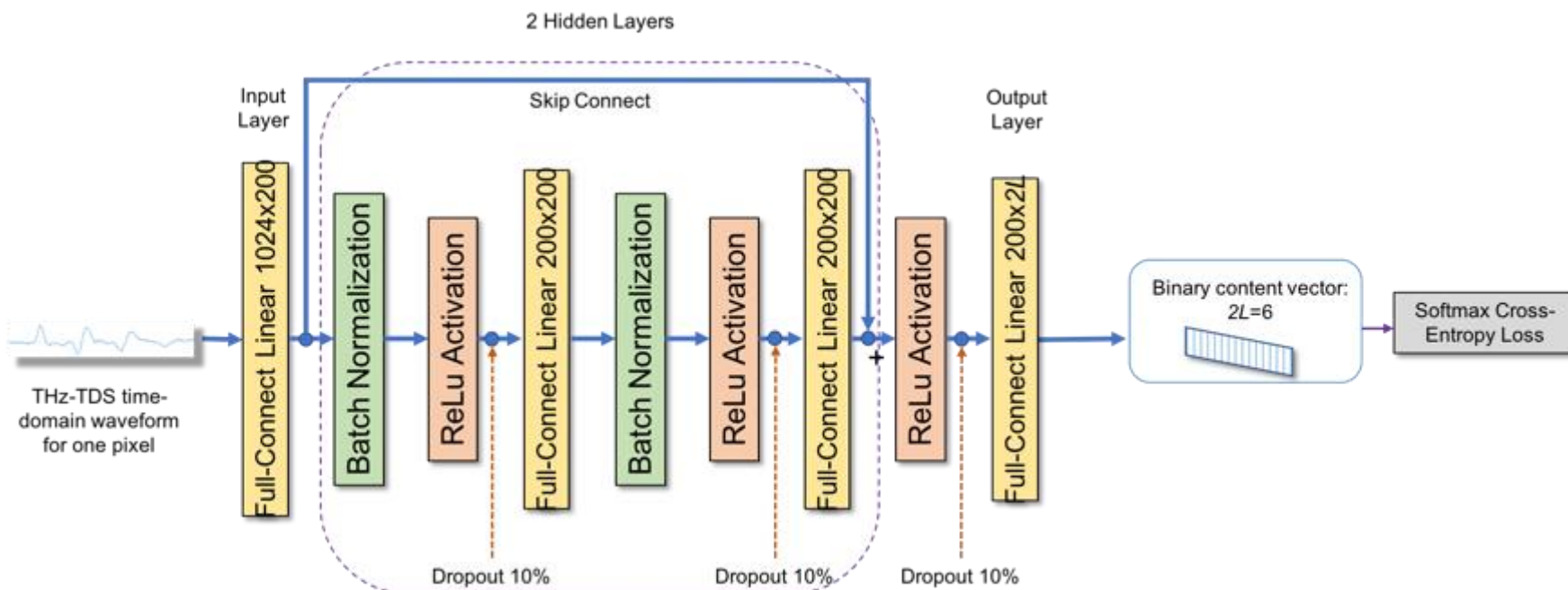
Training



Testing

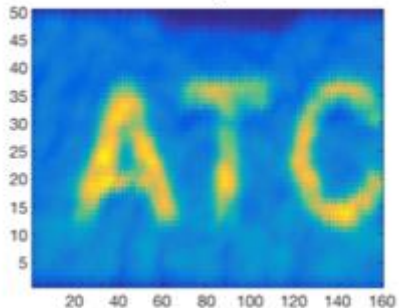


Deep Residual Network (ResNet)

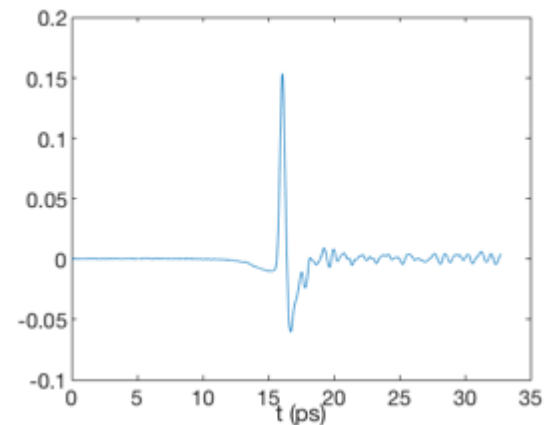
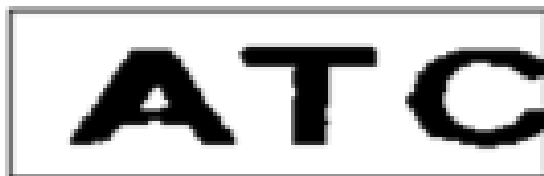


Synthesized Dataset with Sweep Distortion

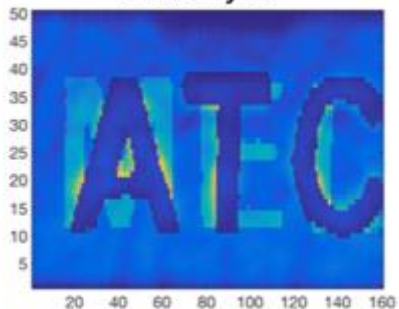
1st layer



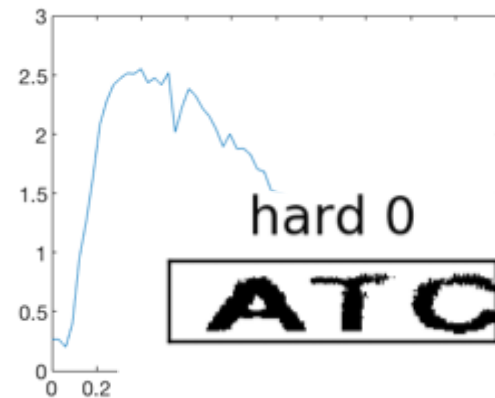
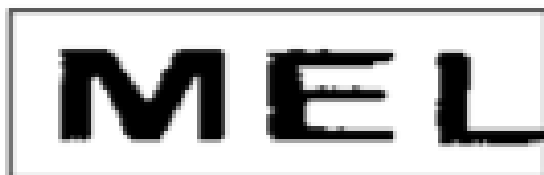
hard 0



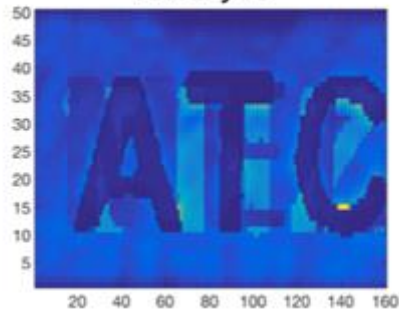
2nd layer



hard 1



3rd layer



hard 2



hard 2



hard 4



