



Depth Correction for Time-of-Flight Camera Using Depth Distortion Dependency on Pulse Width of Irradiated Light

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Tenta. Sasaya*, Wataru. Watanabe, Toshiyuki. Ono

Corporate Research & Development Center, Toshiba Corporation

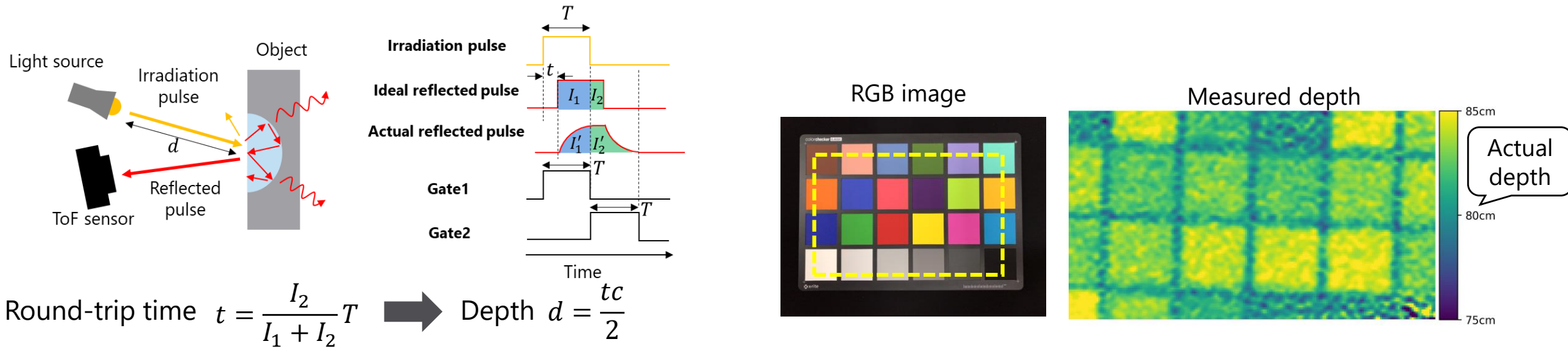
Background

- Time-of-Flight (ToF) Camera

- A kind of 3D camera that estimates depth based on round-trip time of light
- Low cost, high speed, and compact form due to simple measurement system

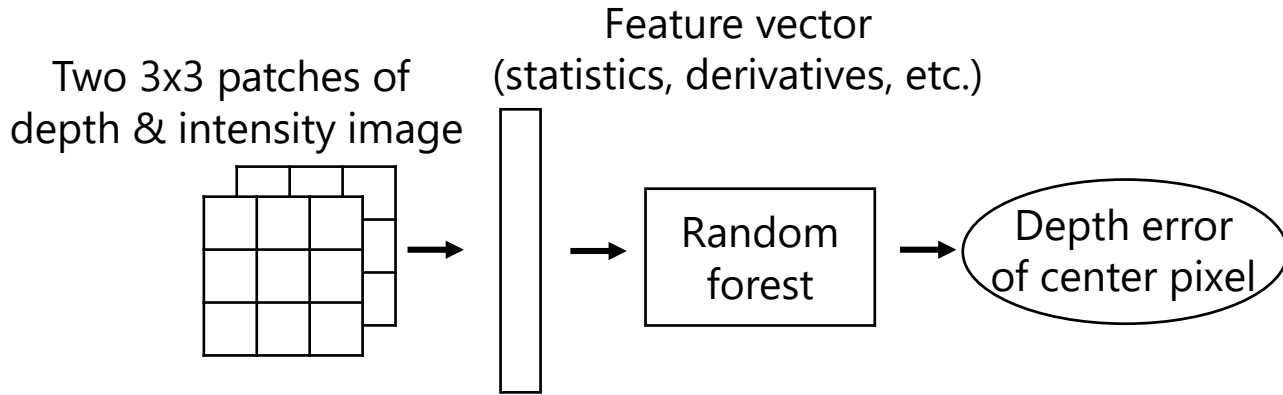
- Drawback: Depth Distortion

- A kind of depth error depending on material of object
- Mainly caused by optical phenomenon (reflection, scattering, transmission, etc.)



**Difficult to model depth distortion process,
so machine learning approach is commonly used for depth correction**

Conventional Depth Correction Method(1): [Fuersattel+,2017]

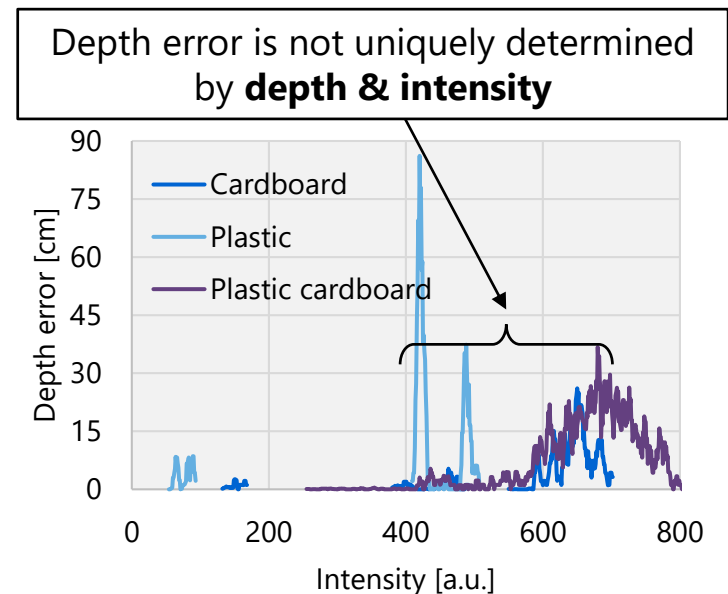
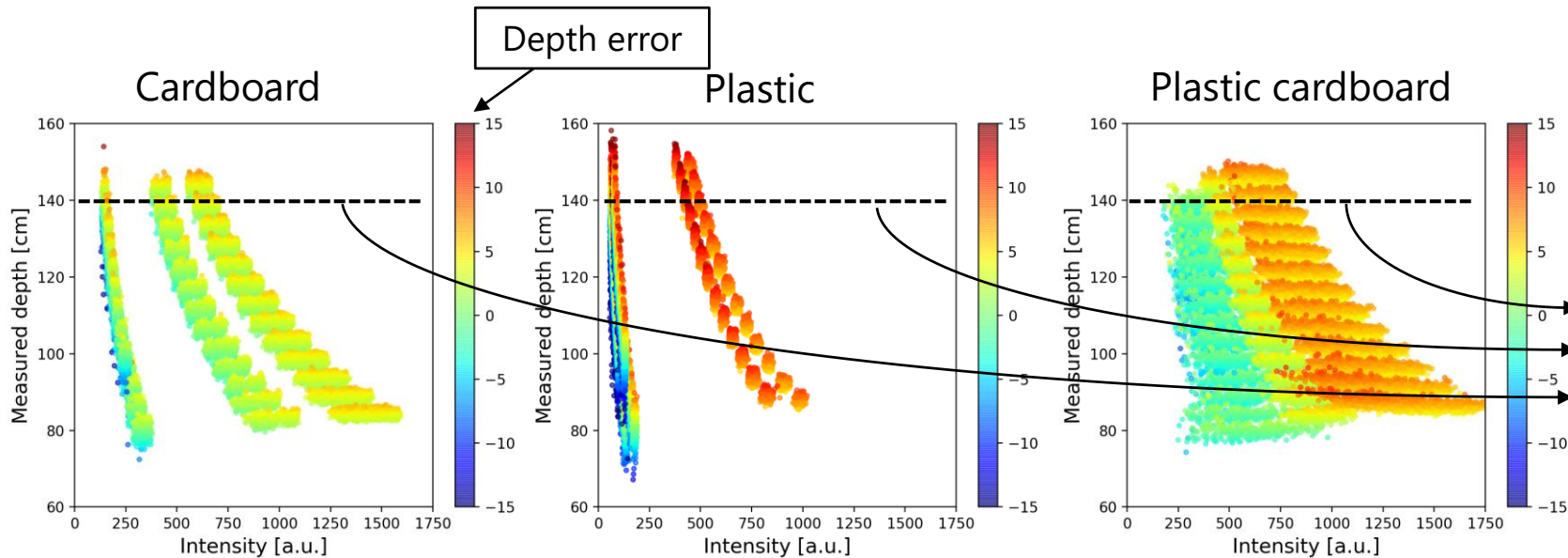


Method

- Calculate feature vector from **depth & intensity**
- Train random forest regressor to estimate depth error

Limitation

- **Doesn't** consider material property explicitly



This method may not work well for diverse materials

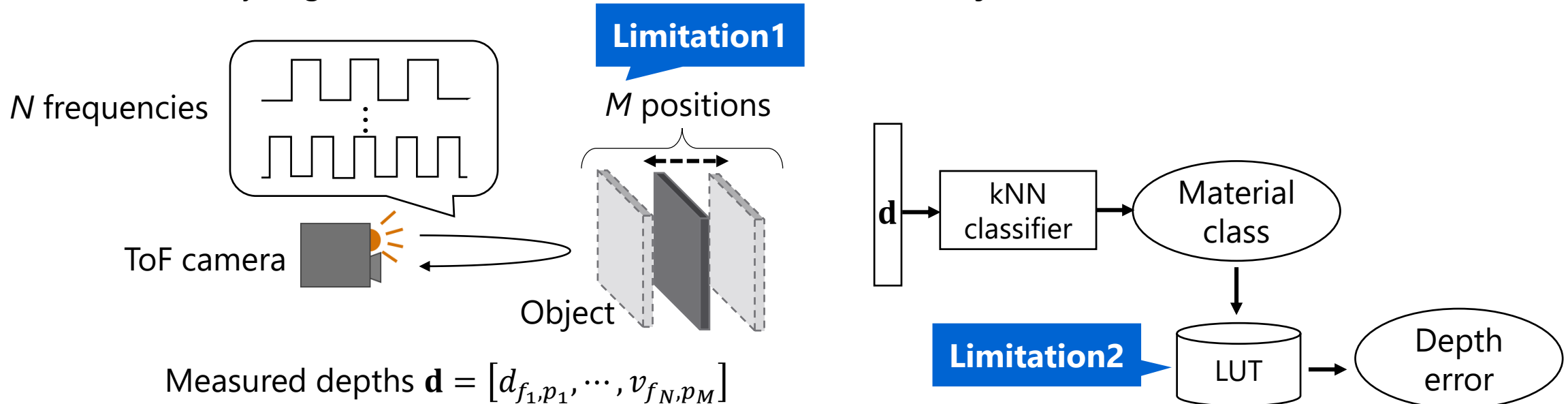
Conventional Depth Correction Method(2): [Tanaka+,2017]

Method

- Measure multiple depths (**not 1-shot**) containing material property using **multiple frequencies & positions**
- Estimate material, then get depth error from LUT using material as query

Limitation

- Require sweeping (**camera and object are fixed** in many industrial applications)
- Performance may degrade for unknown materials (LUT includes **only known materials**)



Proposed method can overcome these limitation because of:

- 1) no sweeping in feature extraction
- 2) direct estimation w/o material classification

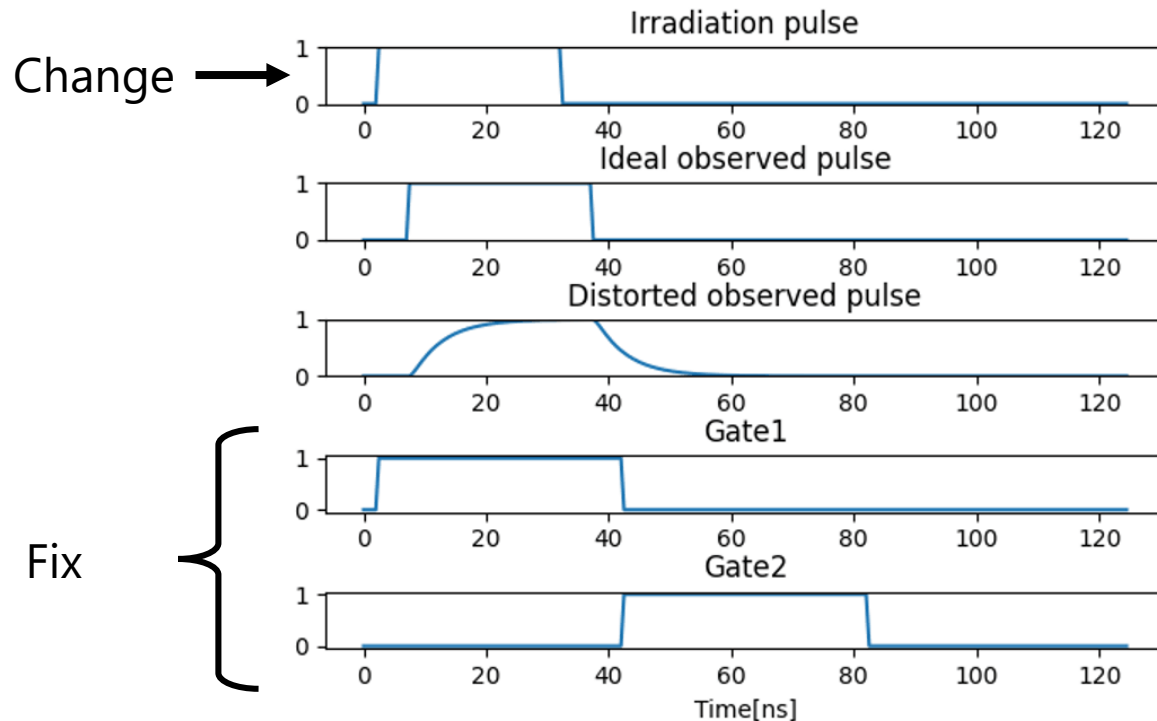
Proposed Method: Feature Extraction

Utilize property that depth changes with pulse width of irradiation light and material

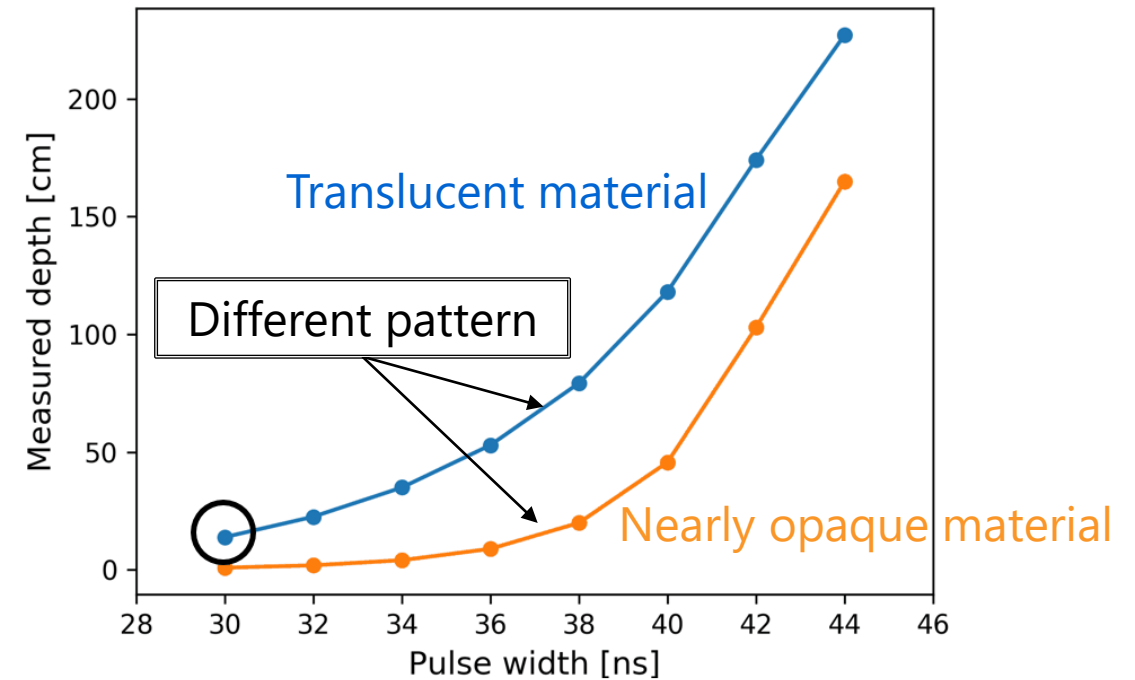
- ✓ Measure multiple depths using **multiple pulse width with fixing Gate1,2** (unusual setting)
- ✓ **No sweeping** unlike conventional method
- ✓ We confirmed proposed **feature vector depends on material** by formula and simulation

Measurement procedure

Pulse width=30



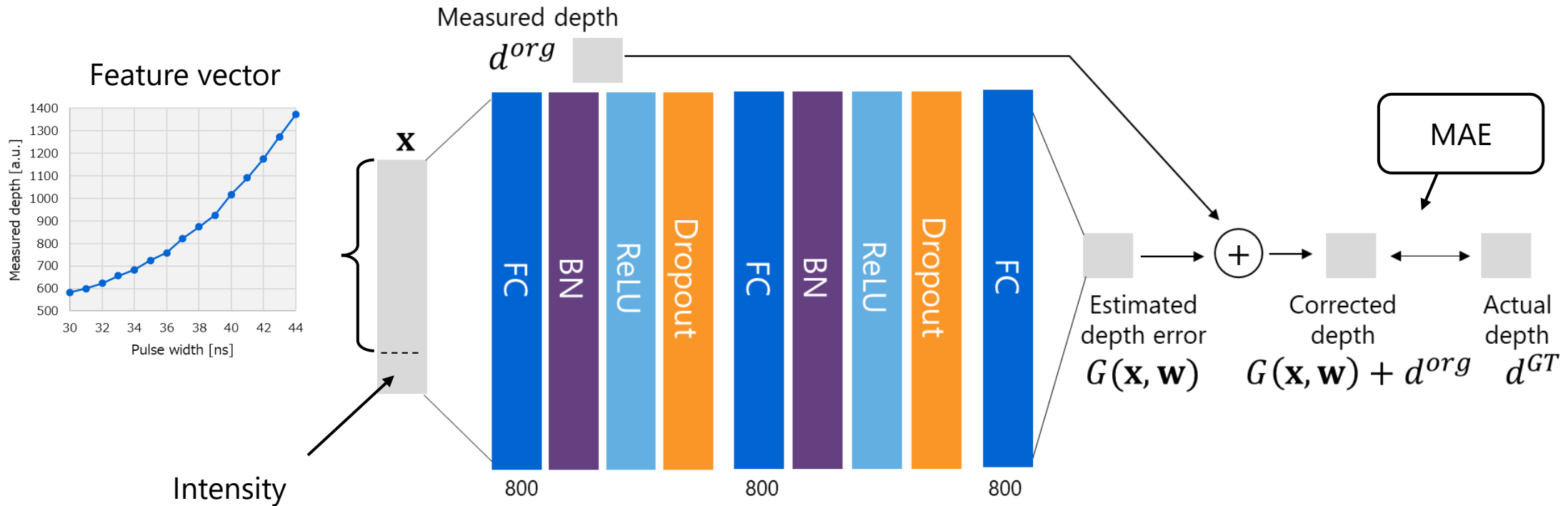
Results of multiple measurements



Proposed Method: Depth Correction

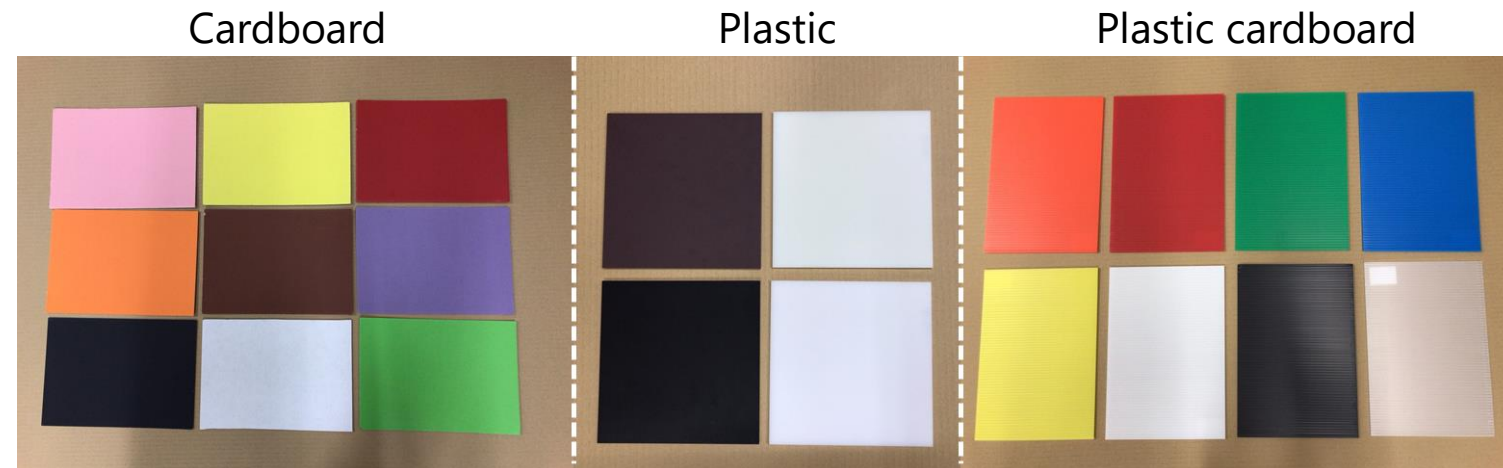
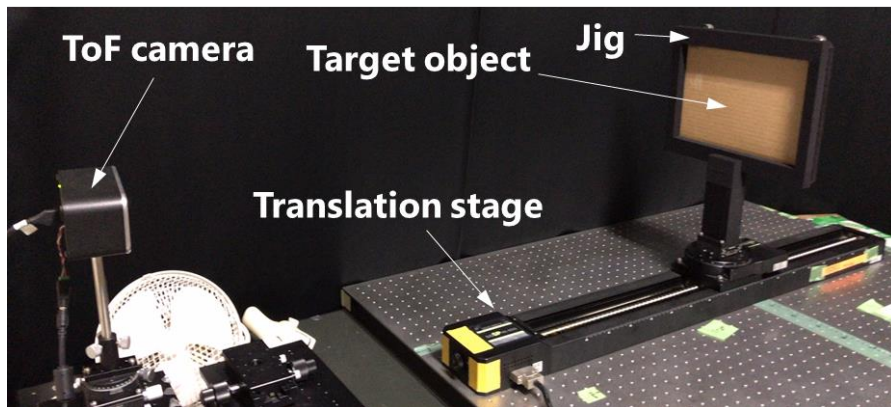
Estimate depth error based on feature vector using neural network

- ✓ Direct manner may deal with **unknown material*** by utilizing similarity of intermediate features
*Not exactly same but similar materials in training data



Experimental Setup

- ToF camera
 - Model: In-house prototype using Panasonic's ToF sensor (MN34902BL)
 - Pulse widths: 30,32,...,44ns (8 pattern)
 - # of frames: 300
 - Calibrate measured value[a.u.] into measured depth[cm] based on depth of reference object
- Object for training data
 - Material: 10 cardboards, 4 plastic boards, 8 plastic cardboards
 - Fixed parallel to sensor surface using photo frame-like jig
 - Distance from ToF camera : 85,90 ,..., 140cm (12 pattern)



Evaluation for Known Materials

- Same material for training and testing
- Randomly selected 80% of pixels for training, remaining 20% for testing
- Calculate MAE of each object, then evaluate $\text{mean}(\text{MAE}_{\text{mean}})$ and $\text{max}(\text{MAE}_{\text{max}})$ for each material

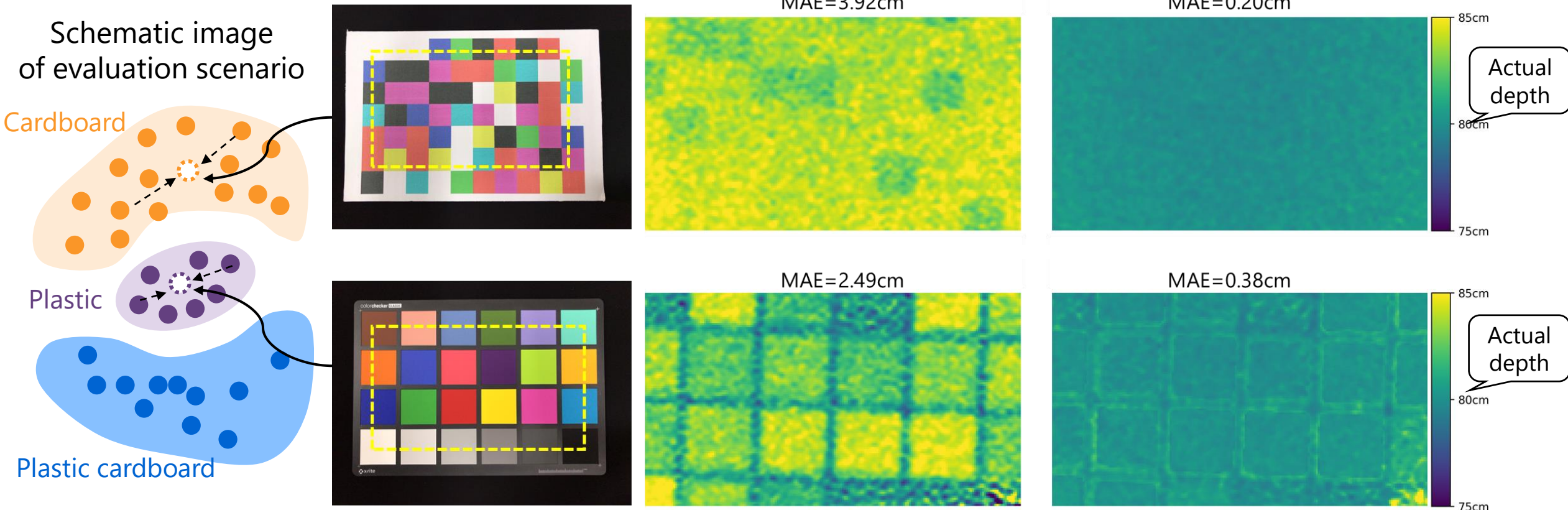
Method	$\text{MAE}_{\text{mean}} / \text{MAE}_{\text{max}} [\text{cm}]$			
	Cardboard	Plastic	Plastic cardboard	All
w/o correction	3.05 / 3.69	6.57 / 9.86	5.25 / 6.83	4.25 / 9.86
[Fuersattel+,2017]	0.55 / 1.60	1.59 / 1.96	0.54 / 0.98	0.73 / 1.96
[Fuersattel+,2017]*	0.71 / 1.37	0.96 / 1.63	0.61 / 1.24	0.70 / 1.63
Ours	0.43 / 0.55	0.94 / 1.19	0.72 / 1.04	0.64 / 1.19

*)Using MLP instead of Random Forest

Proposed method outperform conventional method

Evaluation for Unknown Materials (1)

- These two objects are not exactly same materials in training data, but similar to plastic or cardboard

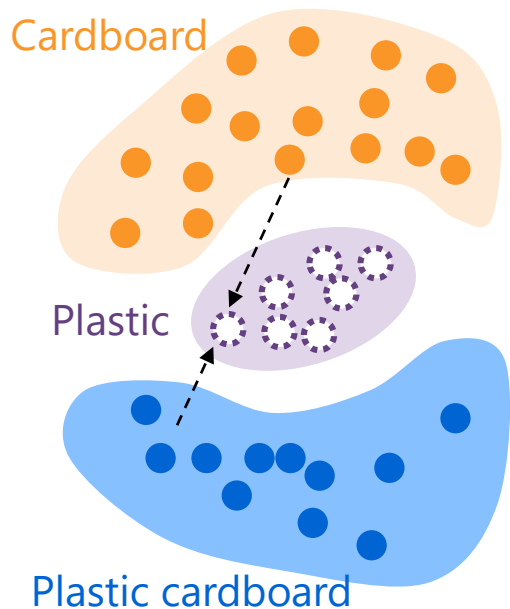


Proposed method work well for slightly different materials from training data

Evaluation for Unknown Materials (2)

- Evaluate unknown materials by changing combination of materials in training data

Schematic image
of evaluation scenario



Material in training data			MAE _{mean} on testing material [cm]		
Cardboard	Plastic	Plastic Cardboard	Cardboard	Plastic	Plastic cardboard
✓			0.24	4.01	2.38
✓	✓		0.33	0.82	2.28
✓		✓	0.33	2.93	0.69

Better MAE despite **unknown material**

Proposed method work well for unknown materials
by interpolating from known material information in the feature space

Summary

◆ Our method

- Feature extraction **without sweeping**
- Direct depth correction **without material classification**

◆ Experimental Results

- Our method can deal with **unknown materials**

◆ Future work

- Expand target materials
- Reduce variation in MAE between materials

TOSHIBA

Thank you

