

## Efficient Multi-Frequency Phase Unwrapping using Kernel Density Estimation

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## Introduction

We introduce an efficient method to unwrap multi-frequency phase estimates for time-offlight ranging. The algorithm generates multiple depth hypotheses and uses a spatial kernel density estimate (KDE) to rank them. We apply the method on the Kinect v2.

- The Kinect v2 is designed for scenes with less than 8m range, but with our method the effective range can be extended.
- When extending the depth range to the maximal value of 18.75m, we get about 52% more valid measurements than existing drivers in *libfreenect2* and *Microsoft Kinect SDK*.

## Results

We apply the method to depth decoding for the Kinect v2 sensor, and compare it to the *Microsoft Kinect SDK* and to the open source driver *libfreenect2*.

- Ground truth is constructed by fusing many depth frames from 9 different camera poses into one very accurate depth image.
- Raw measurements from the ground truth pose are decoded into depth images using *Microsoft*, *libfreenect2* and our method.
- A point is counted as an inlier when a method outputs a depth estimate that correspond to a correct unwrapping, which we set to be within 30cm from the ground truth.
- We evaluate on the **kitchen** dataset with maximal depth of 6.7m and the **lecture** dataset with
- Runs in ≈ 190Hz on a Nvidia GeForce GTX 760 GPU. Code is available at: http://www.cvl.isy.liu.se/research/ datasets/kinect2-dataset/

## Method

In amplitude modulated time-of-flight ranging the depth is obtained by measuring the phase shift  $\phi_m \in [0, 2\pi)$  of the modulated signal given the frequencies  $f_m$ .



maximum depth of 14.6m.



**Figure 2:** Inlier and outlier rate plots and corresponding ground truth depth images. Each point or curve is the average over 25 frames. In the **kitchen** (depth limited) curve the algorithms assume a maximum depth of 8m, which simplifies the outlier rejection.

• A clear improvement can be observed in the depth images, especially in large depth scenes.



**Figure 1:** Left: The Kinect v2 sensor, right: wrapped phases for Kinect v2, in the range 0 to 25 meters. Top to bottom:  $\phi_0$ ,  $\phi_1$ ,  $\phi_2$ . The dashed line at 18.75 meters indicates the common wrap-around point for all three phases. Just before this line we have  $n_0 = 9$ ,  $n_1 = 1$ , and  $n_2 = 14$ .

- The corresponding unwrapped phase measurements are  $\phi_m + 2\pi n_m$ .
- Phase unwrapping means finding the unwrapping coefficients  $\mathbf{n} = (n_0, \dots, n_{M-1}).$
- Each vector **n** corresponds to a hypothesis of the depth  $t^i$ .
- Our method considers several hypotheses for each pixel location and selects the one with the highest kernel density value in a spatial neighbourhood  $\mathcal{N}(\mathbf{x})$ :



**Figure 3:** Single frame output. Left: *libfreenect2*, Center: proposed method. Right: corresponding RGB image. Pixels suppressed by outlier rejection are shown in green.

• Recordings of raw Kinect v2 measurements were unwrapped and passed to the Kinect fusion implementation KinFu in the *Point Cloud Library*.



 $\sum_{j \in \mathcal{I}, k \in \mathcal{N}(\mathbf{x})} w_{jk}$  $w_{ik} = g(\mathbf{x} - \mathbf{x}_k, \sigma) p(t^i | \mathbf{n}_i) p(t^i | \mathbf{a}_i), K(x) = e^{\frac{-x^2}{2h^2}}$ 

The three factors in  $w_{ik}$  are:

- the spatial weight  $g(\mathbf{x} \mathbf{x}_k, \sigma)$ .
- the unwrapping likelihood  $p(t^i(\mathbf{x})|\mathbf{n}_i(\mathbf{x}))$ .
- the phase likelihood  $p(t^i(\mathbf{x})|\mathbf{a}_i(\mathbf{x}))$ , where  $\mathbf{a}_i = (a_0, \dots, a_{M-1})$ , are the amplitudes.

The final hypothesis selection is then made as:

$$i^* = \arg\max_{i\in\mathcal{I}} p(t^i).$$

• Inliers are classified by  $p(t^i) > T$ .

**Figure 4:** KinFu scans of two different scenes using depth images produced by *libfreenect2* and the proposed method. The input duration was 200 frames.