



➤ The homography between the two views can be determined with the plane parameter (n_i, e_i) with known camera poses:

$$H_i(n_i, e_i) \sim K(R - \frac{tn_i^T}{e_i})K^{-1}$$

Key Idea

- Existing learning-based 3D plane reconstruction methods:
 - ❑ rely on single-view regression;
 - ❑ suffer from depth scale ambiguity.
- We propose *the first end-to-end framework* to reconstruct planes from multi-view stereo.
 - ❑ apply **slanted plane hypothesis** to regress planes in plane MVS branch;
 - ❑ decouple the problem into a *plane detection* branch and a *plane MVS* branch, associating them with the proposed loss for joint optimization.

Then we can regress (n_i, e_i) with an MVS framework with a set of **slanted** plane hypothesis.

- For plane detection branch, we adopt state-of-the-art plane detection algorithm, **PlaneRCNN**, to predict plane masks.
- We propose a **soft-pooling loss** to build association between plane detection and plane geometry through the supervision of reconstructed planar depth map for joint optimization.
- For joint optimization on multiple tasks, we further apply the **learnable uncertainty** on different losses, and apply **convex upsampling** to retain the fine-grained geometric details.