# **Co-SLAM Mapping**

```
    ⇔ Status Not started
```

### 初始化

```
self.create_bounds()
self.keyframeDatabase = self.create_kf_database(config)
def create_kf_database(self, config):
        Create the keyframe database
        num_kf = int(self.dataset.num_frames // self.config['mapping']['keyframe_every'] + 1) # 401
        print('#kf:', num_kf)
        print('#Pixels to save:', self.dataset.num_rays_to_save) # 40800 total * 0.05
        return KeyFrameDatabase(config,
                                self.dataset.H,
                                self.dataset.W,
                                num kf.
                                self.dataset.num_rays_to_save,
                                self.device)
\verb|self.model| = \verb|JointEncoding(config, self.bounding_box).to(self.device)| \\
def __init__(self, config, bound_box):
       super(JointEncoding, self).__init__()
        self.config = config
        self.bounding_box = bound_box
        self.get_resolution()
        self.get_encoding(config)
        self.get_decoder(config)
```

## first\_frame\_mapping

```
Step 1: 读取pose
Step 2: 图片上随机采取2048个pixel
Step 3: 得到对应的direction, rgb, depth
Step 4: 根据pose得到对应的rays_o. rays_d
Step 5: 根据depth值在表面随机采样11个z_vals, 在free space采样32个z_vals
Step 6: 将采样点转换为实际的世界坐标系下的坐标
Step 7: 将采样点输入网络得到预测值 [rgb, sdf]
Step 8: rendering
```

### 模型训练

Co-SLAM Mapping 1

```
raw: [N_rays, N_samples, 4]
inputs_flat = torch.reshape(query_points, [-1, query_points.shape[-1]])

embed = self.embed_fn(inputs_flat) # hash encoding
embe_pos = self.embedpos_fn(inputs_flat) # one-blob
if not self.config['grid']['oneGrid']:
    embed_color = self.embed_fn_color(inputs_flat)
    return self.decoder(embed, embe_pos, embed_color)
return self.decoder(embed, embe_pos)
```

## rendering

```
rgb_map, disp_map, acc_map, weights, depth_map, depth_var = self.raw2outputs(raw, z_vals, self.config['training']['white_bkgd'])
```

```
\label{lem:def_sdf2} \mbox{def sdf2weights(self, sdf, z_vals, args=None):}
                                 Convert signed distance function to weights.
                                 Params:
                                                 sdf: [N_rays, N_samples]
                                                   z_vals: [N_rays, N_samples]
                                weights: [N_rays, N_samples]
                                 Returns:
                                 # 计算初始权重
                                 weights = torch.sigmoid(sdf / args['training']['trunc']) * torch.sigmoid(-sdf / args['training']['trunc']] * torch.sigmoid(-sdf / args['training']['training']['training']['training'] * torch.sigmoid(-sdf / args['training']['training']['training']['training']['training']['training'] * torch.sigmoid(-sdf / args['training']['training']['training']['training']['training']['training'] * torch.sigmoid(-sdf / args['training']['training']['training']['training']['training']['training']['training']['training']['training']['training']['training']['
                                 # 计算哪些相邻的SDF值之间发生了符号的变化
                                 signs = sdf[:, 1:] * sdf[:, :-1]
                                 mask = torch.where(signs < 0.0, torch.ones_like(signs), torch.zeros_like(signs))</pre>
                                 # 找到符号变化的位置: argmax即每一条光线上首次符号变化的位置,即第一个表面的位置
                                 inds = torch.argmax(mask, axis=1)
                                 inds = inds[..., None]
                                 # 获取第一个表面的z值
                                 z_{min} = torch.gather(z_vals, 1, inds) # The first surface
                                 mask = torch.where(z\_vals < z\_min + args['data']['sc\_factor'] * args['training']['trunc'], torch.ones\_like(z\_vals), torch.zeros\_like(z\_vals), torc
                                 weights = weights * mask
                                 return weights / (torch.sum(weights, axis=-1, keepdims=True) + 1e-8)
```

Co-SLAM Mapping 2