

Geometric Computer Vision (HW)

Denis Rollov.

```
# TODO: write your code to construct a world-frame point cloud from a depth image,  
# using known intrinsic and extrinsic camera parameters.  
# Hints: use the class `RaycastingImaging` to transform image to points in camera frame,  
# use the class `CameraPose` to transform image to points in world frame.
```

In this #TODO we need to calculate the following variables: `points_i`, `pose_i`, `imaging_i` (not incorporate `image_i`, `distances_i`).

In order to transform image to points in world frame, we merely need to create object `CameraPose`. To transform image to points in camera frame we need to define `RaycastingImaging` via two variables: **resolution_image** (it is just the resolution of our image (`shape[0]`)) and **resolution_3d** (mm/pixel) as well. In order to calculate **points_i** initially it is important to transform **images_i** to points and then convert the points from the camera to world. Subsequently, we will be able to construct a world-frame point cloud from a depth image.

```
# Reproject points from view_j to view_i, to be able to interpolate in view_i.  
# We are using parallel projection so this explicitly computes  
# (u, v) coordinates for reprojected points (in image plane of view_i).  
# TODO: your code here: use functions from CameraPose class  
# to transform `points_j` into coordinate frame of `view_i`
```

Simple step: create a transformation from world to camera via method **world_to_camera**.

```
# For each reprojected point, find K nearest points in view_i,  
# that are source points/pixels to interpolate from.  
# We do this using imaging_i.rays_origins because these  
# define (u, v) coordinates of points_i in the pixel grid of view_i.  
# TODO: your code here: use cKDTree to find k=`nn_set_size` indexes of  
# nearest points for each of points from `reprojected_j`
```

In order to realize this part of the code it is necessary to find the solution what to input into **cKDTree**. The input is x, y coordinates of **rays_origins**. After that it is necessary to query the kd-tree for nearest neighbors (inputs in query are reprojected x and y and number of the nearest neighbors to return).

```
# Build an [n, 3] array of XYZ coordinates for each reprojected point by taking
# UV values from pixel grid and Z value from depth image.
# TODO: your code here: use `point_nn_indexes` found previously
```

Before this step it is important to initialize flattened vectors of image and distance (in order to speed-up our code and not repeat the same calculations). So, I just took number of neighbours indexes and concat x,y (array $R^{n, 2}$) and z (column-vector).

```
# TODO: compute a flag indicating the possibility to interpolate
# by checking distance between `point_from_j` and its `point_from_j_nns`
# against the value of `distance_interpolation_threshold`
```

In order to calculate the distance it is necessary to use Euclidean distance. The next step is applying mask into our obtained distances.

```
# TODO: your code here: use `interpolate.interp2d`
# to construct a bilinear interpolator from distances predicted
# in `view_i` (i.e. `distances_i`) into the point in `view_j`.
# Use the interpolator to compute an interpolated distance value.
```

Bilinear interpolator from predicted distance can be constructed via three variables: x-values, y-values, flatten vector of number of corresponding number of neighbours. Subsequently, interpolator just takes two values as input: xnew and ynew from reprojected matrix (transformation world to camera).