

4. IMPLEMENTASI SISTEM

Pada bab ini akan membahas implementasi dari desain sistem yang telah dijelaskan pada bab sebelumnya.

Tabel 4.1 Hubungan Segmen Program dengan Desain Sistem

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4.1 Konversi *Dataset*

Berikut adalah segmen program yang berhubungan dengan melakukan konversi dataset dari format OBJ menjadi GLB.

Segmen Program 4.1 Konversi *Dataset*

```
import os
import sys
import json
import shutil
import subprocess
from tqdm.auto import tqdm
taxonomy_list = json.loads(open("shapenetcore.taxonomy.json").read())
folder_dict = {}
for taxon in taxonomy_list:
    folder_dict[taxon["metadata"]["name"]] =
    taxon["metadata"]["label"].split(",")[0]

    if "children" in taxon:
        for taxon_child in taxon["children"]:
```

```

        folder_dict[taxon_child["metadata"]["name"]] =
taxon_child["metadata"]["label"].split(",")[0]

        if "children" in taxon_child:
            for taxon_child_child in taxon_child["children"]:
                folder_dict[taxon_child_child["metadata"]["name"]] =
taxon_child_child["metadata"]["label"].split(",")[0]

                if "children" in taxon_child_child:
                    for taxon_child_child_child in
taxon_child_child["children"]:

folder_dict[taxon_child_child_child["metadata"]["name"]] =
taxon_child_child_child["metadata"]["label"].split(",")[0]

                    if "children" in taxon_child_child_child:
                        for taxon_child_child_child_child in
taxon_child_child_child["children"]:

folder_dict[taxon_child_child_child_child["metadata"]["name"]] =
taxon_child_child_child_child["metadata"]["label"].split(",")[0]

                        if "children" in
taxon_child_child_child_child:
                            for
taxon_child_child_child_child_child in
taxon_child_child_child_child["children"]:

folder_dict[taxon_child_child_child_child_child["metadata"]["name"]] =
taxon_child_child_child_child_child["metadata"]["label"].split(",")[0]
target_cat_path_list = []
obj_path_list = []
for folder_path in os.listdir("./ShapeNetCore"):

target_cat_path_list.append(f"./ShapeNetCoreGLB/{folder_dict[str(folder_path)]}
}")
    for cat_path in os.listdir(f"./ShapeNetCore/{folder_path}"):
        source_obj_path =
os.path.abspath(f"./ShapeNetCore/{folder_path}/{cat_path}/models/model_normali
zed.obj")
        target_glb_path =
os.path.abspath(f"./ShapeNetCoreGLB/{folder_dict[str(folder_path)]}/{cat_path}
.glb")
        obj_path_list.append((source_obj_path, target_glb_path))
if os.path.exists("./ShapeNetCoreGLB/"):
    shutil.rmtree("./ShapeNetCoreGLB/")
os.mkdir("./ShapeNetCoreGLB/")
for new_dir in target_cat_path_list:
    os.mkdir(new_dir)
for curr_source, curr_target in tqdm(obj_path_list):
    subprocess.Popen(

```

```

        [
            'obj2gltf',
            '-i',
            f'{curr_source}',
            '-o',
            f'{curr_target}'
        ],
        shell=True,
        stderr=subprocess.STDOUT,
        universal_newlines=True,
    )
missing_list = []
for curr_source, curr_target in tqdm(obj_path_list):
    if not os.path.exists(curr_target):
        missing_list.append((curr_source, curr_target))
for curr_source, curr_target in tqdm(missing_list):
    if os.path.exists(curr_source):
        subprocess.check_output(
            [
                'obj2gltf',
                '-i',
                f'{curr_source}',
                '-o',
                f'{curr_target}'
            ],
            shell=True,
            stderr=subprocess.STDOUT,
            universal_newlines=True,
        )
missing_list_extended = []
for curr_source, curr_target in tqdm(obj_path_list):
    if not os.path.exists(curr_target):
        missing_list_extended.append((curr_source, curr_target))

```

4.2 Preprocessing

Berikut adalah segmen-segmen program yang berhubungan dengan *preprocessing dataset* sebelum menjadi input model.

Segmen Program 4.2 *Rendering Image*

```

import pyrender
import os
import trimesh
import numpy as np
import matplotlib.pyplot as plt
import numpy as np
import scipy

```

```

from scipy.io import savemat
from scipy.spatial.transform import Rotation
from sklearn.utils import check_random_state
from tqdm import tqdm
from trimesh.transformations import rotation_matrix, quaternion_from_matrix,
quaternion_matrix
import cv2
import PIL
from transforms3d.quaternions import quat2mat

rotation_angle_x = np.deg2rad(i * ROTATION_STEP)
rotation_matrix_x = Rotation.from_rotvec(rotation_angle_x * np.array([0, 1,
0])).as_matrix()

diagonal_angle_y = np.deg2rad(DIAGONAL_ANGLE)
diagonal_matrix_y = Rotation.from_rotvec(diagonal_angle_y * np.array([0, 1,
0])).as_matrix()

top_down_angle_x = np.deg2rad(TOP_DOWN_ANGLE)
top_down_matrix_x = Rotation.from_rotvec(top_down_angle_x * np.array([1, 0,
0])).as_matrix()

combined_rotation_matrix = np.dot(top_down_matrix_x, np.dot(diagonal_matrix_y
rotation_matrix_x))

# Set the object's pose with combined rotation
object_pose = np.eye(4)
object_pose[:3, :3] = combined_rotation_matrix
object_pose[:3, 3] = object_position

scene = pyrender.Scene()
# Create a camera
camera = pyrender.OrthographicCamera(xmag=0.5, ymag=0.5, znear=0.5, zfar=1.3)
camera_pose = np.eye(4)
camera_pose[:3, 3] = [0, 0, 1]
# Camera pose is not important for this example
scene.add(camera, pose=camera_pose)

# Create a light

```

```

light = pyrender.DirectionalLight(color=np.ones(3), intensity=2.0)
# Light pose is not important for this example
scene.add(light, pose=np.eye(4))

# Add the rotated mesh to the scene
scene.add(mesh, pose=object_pose)

# Render the scene
renderer_input = pyrender.OffscreenRenderer(
    viewport_width=512, viewport_height=512)
color_input, depth_input = renderer_input.render(scene)

# Explicitly release resources associated with the mesh
renderer_input.delete()

```

Segmen Program 4.3 *Cropping Image*

```

y_input_image = cv2.cvtColor(color_input, cv2.COLOR_BGR2GRAY)

# To find upper threshold, we need to apply Otsu's thresholding
upper_threshold, thresh_input_image = cv2.threshold(
    gray_input_image, thresh=0, maxval=255, type=cv2.THRESH_BINARY +
    cv2.THRESH_OTSU)

# Calculate lower threshold
lower_threshold = 0.5 * upper_threshold

# Apply canny edge detection
canny = cv2.Canny(color_input, lower_threshold, upper_threshold)
# Finding the non-zero points of canny
pts = np.argwhere(canny > 0)

# Finding the min and max points
try:
    y1, x1 = pts.min(axis=0)
    y2, x2 = pts.max(axis=0)

    # Crop ROI from the givn image
    output_image = color_input[y1:y2, x1:x2]

```

Segmen Program 4.4 *Resizing dan Padding Image*

```
old_size = output_image.shape[:2]
desired_img_size = 64

ratio_img = float(desired_img_size)/max(old_size)
new_img_size = tuple([int(x * ratio_img) for x in old_size])

im = cv2.resize(output_image, (new_img_size[1], new_img_size[0]))

delta_w_img = desired_img_size - new_img_size[1]
delta_h_img = desired_img_size - new_img_size[0]
top_img, bottom_img = delta_h_img//2, delta_h_img-(delta_h_img//2)
left_img, right_img = delta_w_img//2, delta_w_img-(delta_w_img//2)

new_im = cv2.copyMakeBorder(im, top_img, bottom_img, left_img, right_img,
                             cv2.BORDER_CONSTANT,
                             value=[255, 255, 255])
```

Segmen Program 4.5 *Rendering Depth Map*

```
quat_matrix = trans_file[j]
print(trans_file[j])
quat2rot_matrix = quat2mat(quat_matrix)

rotation_angle_z = np.deg2rad(-90)
rotation_axis_z = [0, 0, 1]
rotation_matrix_z = Rotation.from_rotvec(rotation_angle_z *
np.array(rotation_axis_z)).as_matrix()

rotation_angle_x = np.deg2rad(-90)
rotation_axis_x = [0, 1, 0]
rotation_matrix_x = Rotation.from_rotvec(rotation_angle_x *
np.array(rotation_axis_x)).as_matrix()

# Combine the rotation matrices
```

```

combined_rotation_matrix = np.dot(quat2rot_matrix, np.dot(rotation_matrix_x,
rotation_matrix_z))

object_pose = np.eye(4)
object_pose[:3, :3] = combined_rotation_matrix[:3, :3]

scene = pyrender.Scene()
# Create a camera
camera = pyrender.OrthographicCamera(xmag=0.5, ymag=0.5, znear=0.1, zfar=1.3)

camera_pose = np.eye(4)
camera_pose[:3, 3] = [0, 0, 1]
scene.add(camera, pose=camera_pose)

# Create a light
light = pyrender.DirectionalLight(color=np.ones(3), intensity=2.0)
scene.add(light, pose=np.eye(4))

# Add the rotated mesh to the scene
scene.add(mesh, pose=object_pose)

# Render the scene
renderer_input = pyrender.OffscreenRenderer(
viewport_width=512, viewport_height=512)
color_input, depth_input = renderer_input.render(scene)

# Explicitly release resources associated with the mesh
renderer_input.delete()

```

Segmen Program 4.6 *Cropping Depth Map*

```

gray_input_image = cv2.cvtColor(color_input, cv2.COLOR_BGR2GRAY)

# To find upper threshold, we need to apply Otsu's thresholding
upper_threshold, thresh_input_image = cv2.threshold(
gray_input_image, thresh=0, maxval=255, type=cv2.THRESH_BINARY + cv2.THRESH_OTSU
)

# Calculate lower threshold
lower_threshold = 0.5 * upper_threshold

```

```

# Apply canny edge detection
canny = cv2.Canny(color_input, lower_threshold, upper_threshold)

# Finding the non-zero points of canny
pts = np.argwhere(canny > 0)

# Finding the min and max points
try:
    y1, x1 = pts.min(axis=0)
    y2, x2 = pts.max(axis=0)

    # Crop ROI from the given image
    output_depth = depth_input[y1:y2, x1:x2]

```

Segmen Program 4.7 *Resizing dan Padding Depth Map*

```

old_size = output_depth.shape[:2]
desired_dp_size = 128

ratio_dp = float(desired_dp_size)/max(old_size)
new_dp_size = tuple([int(x * ratio_dp) for x in old_size])

dp = cv2.resize(output_depth, (new_dp_size[1], new_dp_size[0]))

delta_w_dp = desired_dp_size - new_dp_size[1]
delta_h_dp = desired_dp_size - new_dp_size[0]
top_dp, bottom_dp = delta_h_dp//2, delta_h_dp-(delta_h_dp//2)
left_dp, right_dp = delta_w_dp//2, delta_w_dp-(delta_w_dp//2)

new_depth = cv2.copyMakeBorder(dp, top_dp, bottom_dp, left_dp, right_dp,
cv2.BORDER_CONSTANT, value=[0, 0, 0])

```

Segmen Program 4.8 *Parameter Tuning dan Pembuatan Mask*

```

depth = np.load(
    f"{self.cfg.path}/depth/{classname}/{filename}.npy")

```

```

mask = np.array([value != 0 for value in depth])
mask = mask.astype(bool)
depth = depth.transpose(1, 2, 0)
depth = (depth - np.min(depth, axis=(0, 1))) / \
        (np.max(depth, axis=(0, 1)) - np.min(depth, axis=(0, 1)))
depth = depth.transpose(2, 0, 1)

depth[depth == 0] = 0.5

background_mask = (depth == 0)
min_val = np.min(depth[depth != 0])
max_val = np.max(depth[depth != 0])

# Generate random noise values for the background pixels
noise_range = (min_val, max_val) # Adjust the range as needed
noise_values = np.random.uniform(
noise_range[0], noise_range[1], size=depth.shape)

# Replace the background pixels with the generated noise values
depth_with_noise = depth.copy()
depth_with_noise[background_mask] = noise_values[background_mask]

```

Segmen Program 4.9 Pembuatan *Ground Truth Point Cloud*

```

def create_dense_point_cloud(vertices, faces, density=1000, random_state=None):
    """
    Create a dense point cloud by subdividing faces and sampling points within
    them.

    Parameters:
        vertices (numpy.ndarray): 2D array of vertex coordinates (n_vertices x
        3).
        faces (numpy.ndarray): 2D array of faces (n_faces x 3).
        density (int): Number of points to sample per unit area.
        random_state (int or RandomState, optional): Seed or RandomState for
        reproducibility.

    Returns:
        numpy.ndarray: Dense point cloud coordinates (n_points x 3).
    """

```

```

"""
# Ensure vertices and faces are NumPy arrays
vertices = np.array(vertices)
faces = np.array(faces)

# Subdivide faces and sample points within each triangle
random_state = check_random_state(random_state)
mesh = trimesh.Trimesh(vertices=vertices, faces=faces)

# Calculate the total area of the triangles in the mesh
total_area = np.sum(trimesh.triangles.area(mesh.triangles))

# Calculate the number of points to sample
num_points = int(density * total_area)

dense_point_cloud, _ = trimesh.sample.sample_surface(
    mesh,
    num_points,
    seed=random_state.randint(2**31)
)

return dense_point_cloud

def visualize_point_cloud(point_cloud):
    """
    Visualize a point cloud using matplotlib.

    Parameters:
        point_cloud (numpy.ndarray): Dense point cloud coordinates (n_points x
3).
    """
    fig = plt.figure()
    ax = fig.add_subplot(111, projection='3d')

    # Extract x, y, and z coordinates from the point cloud
    x = point_cloud[:, 0]
    y = point_cloud[:, 1]
    z = point_cloud[:, 2]

```

```

    ax.scatter(x, y, z, s=1, c='b', marker='o') # Adjust marker size (s) as
needed

    ax.set_xlabel('X')
    ax.set_ylabel('Y')
    ax.set_zlabel('Z')

plt.show()

```

4.3 Model Generasi *Point Cloud* dan Implementasi *Fusing*

Berikut adalah segmen-segmen program yang berkaitan dengan arsitektur model serta proses *post processing*, yaitu implementasi *fusing* dengan *invers matriks*.

Segmen Program 4.10 Model Generasi *Point Cloud*

```

import torch
from torch import nn
from torch.nn import functional as F
import config

def conv2d_block(in_c, out_c):
    return nn.Sequential(
        nn.Conv2d(in_c, out_c, 3, stride=2, padding=1),
        nn.BatchNorm2d(out_c),
        nn.ReLU(),
    )

def deconv2d_block(in_c, out_c):
    return nn.Sequential(
        nn.Conv2d(in_c, out_c, 3, stride=1, padding=1),
        nn.BatchNorm2d(out_c),
        nn.ReLU(),
    )

def linear_block(in_c, out_c):

```

```

return nn.Sequential(
    nn.Linear(in_c, out_c),
    nn.BatchNorm1d(out_c),
    nn.ReLU(),
)

def pixel_bias(inputViewN, outW, outH, renderDepth):
    X, Y = torch.meshgrid([torch.arange(outH), torch.arange(outW)])
    X, Y = X.float(), Y.float() # [H,W]
    initTile = torch.cat([
        X.repeat([inputViewN, 1, 1]), # [V,H,W]
        Y.repeat([inputViewN, 1, 1]), # [V,H,W]
        torch.ones([inputViewN, outH, outW]).float() * renderDepth,
        torch.zeros([inputViewN, outH, outW]).float(),
    ], dim=0) # [4V,H,W]

    return initTile.unsqueeze_(dim=0) # [1,4V,H,W]

class Encoder(nn.Module):
    """Encoder of Structure Generator"""

    def __init__(self, inputViewN):
        super(Encoder, self).__init__()

        self.conv1 = conv2d_block(3*inputViewN, 96)
        self.conv2 = conv2d_block(96, 128)
        self.conv3 = conv2d_block(128, 192)
        self.conv4 = conv2d_block(192, 256)
        self.fc1 = linear_block(4096, 2048) # After flatten
        self.fc2 = linear_block(2048, 1024)
        self.fc3 = nn.Linear(1024, 512)

    def forward(self, x):
        x = self.conv1(x)
        x = self.conv2(x)
        x = self.conv3(x)
        x = self.conv4(x)

```

```

        x = self.fc1(x.reshape(-1, 4096))
        x = self.fc2(x)
        x = self.fc3(x)

    return x

class Decoder(nn.Module):
    """Build Decoder"""

    def __init__(self, inputViewN, outW, outH, renderDepth):
        super(Decoder, self).__init__()
        self.inputViewN = inputViewN

        self.relu = nn.ReLU()
        self.fc1 = linear_block(512, 1024)
        self.fc2 = linear_block(1024, 2048)
        self.fc3 = linear_block(2048, 4096)
        self.deconv1 = deconv2d_block(256, 192)
        self.deconv2 = deconv2d_block(192, 128)
        self.deconv3 = deconv2d_block(128, 96)
        self.deconv4 = deconv2d_block(96, 64)
        self.deconv5 = deconv2d_block(64, 48)
        self.pixel_conv = nn.Conv2d(48, inputViewN*4, 1, stride=1, bias=False)
        self.pixel_bias = pixel_bias(inputViewN, outW, outH, renderDepth)

    def forward(self, x):
        x = self.relu(x)
        x = self.fc1(x)
        x = self.fc2(x)
        x = self.fc3(x)
        x = x.reshape([-1, 256, 4, 4])
        x = self.deconv1(F.interpolate(x, scale_factor=2))
        x = self.deconv2(F.interpolate(x, scale_factor=2))
        x = self.deconv3(F.interpolate(x, scale_factor=2))
        x = self.deconv4(F.interpolate(x, scale_factor=2))
        x = self.deconv5(F.interpolate(x, scale_factor=2))
        x = self.pixel_conv(x) + self.pixel_bias.to(x.device)
        XYZ, maskLogit = torch.split(

```

```

        x, [self.inputViewN * 3, self.inputViewN], dim=1)

    return XYZ, maskLogit

class Generator(nn.Module):
    def __init__(self, encoder=None, decoder=None, inputViewN=4,
                  outW=128, outH=128, renderDepth=1.0):
        super(Generator, self).__init__()

        if encoder:
            self.encoder = encoder
        else:
            self.encoder = Encoder(inputViewN)

        if decoder:
            self.decoder = decoder
        else:
            self.decoder = Decoder(inputViewN, outW, outH, renderDepth)

    def forward(self, x):
        latent = self.encoder(x)
        XYZ, maskLogit = self.decoder(latent)

        return XYZ, maskLogit

class Discriminator(nn.Module):
    def __init__(self, inputViewN=8):
        super(Discriminator, self).__init__()
        self.main = nn.Sequential(
            # input is `(nc) x 64 x 64`
            nn.Conv2d(inputViewN, 128, 3, 2, 1, bias=False),
            nn.LeakyReLU(0.2, inplace=True),
            # nn.Dropout(p=0.2),

            # state size. `(ndf) x 32 x 32`
            nn.Conv2d(128, 256 * 2, 3, 2, 1, bias=False),
            nn.BatchNorm2d(256 * 2),

```

```

        nn.LeakyReLU(0.2, inplace=True),
        # nn.Dropout(p=0.2),

        # state size. ``(ndf*2) x 16 x 16``
        nn.Conv2d(256 * 2, 512 * 4, 3, 2, 1, bias=False),
        nn.BatchNorm2d(512 * 4),
        nn.LeakyReLU(0.2, inplace=True),
        # nn.Dropout(p=0.2),

        # state size. ``(ndf*4) x 8 x 8``
        nn.Conv2d(512 * 4, 1024 * 8, 3, 2, 1, bias=False),
        # nn.BatchNorm2d(1024 * 8),
        # nn.LeakyReLU(0.2, inplace=True),

        # state size. ``(ndf*8) x 4 x 4``
        # nn.Conv2d(1024 * 8, 1, 3, 1, 0, bias=False),
        nn.Sigmoid()
    )

    def forward(self, input):
        return self.main(input)

if __name__ == '__main__':
    import config
    cfg = config.get_arguments()
    encoder = Encoder(cfg.inputViewN)
    decoder = Decoder(cfg.inputViewN, cfg.outW, cfg.outH, cfg.renderDepth)

    generator = Generator()
    discriminator = Discriminator()

```

Segmen Program 4.11 *Post Processing Fusion*

```

import numpy as np
import torch
import torch.nn as nn
import torch.nn.functional as F

```

```

from trimesh.transformations import quaternion_from_matrix, quaternion_matrix,
inverse_matrix, translation_matrix, rotation_matrix

def fuse3D(cfg, XYZ, maskLogit, fuseTrans, chunkSize):
    """Fuse multiple depth views into a 3D point cloud representation
    Args:
        output of structure generator
            XYZ (tensor:[B,3V,H,W]): x,z,y of V different fixed views
            maskLogit (tensor:[B,V,H,W]): mask of V different fixed views
        output of render module
            fuseTrans (Tensor:[V, 4])
    Return:
        XYZid (Tensor [B,3,VHW]): point clouds
        ML (Tensor [B,1,VHW]): depth stack
    """
    t_view = torch.Tensor([1, 1, 0])

    # t_view = torch.Tensor([1, 1, 0])

    # t_view = torch.Tensor([1, 0, 0])

    transMat = translation_matrix(t_view)
    transMat = transMat[None, ...] # expand to 1,4,4
    transMat = np.repeat(transMat, cfg.inputViewN,
                        axis=0) # repeat 1 to n views
    transMat = torch.from_numpy(transMat).float().to(cfg.device)

    q_view = fuseTrans # [V, 4]

    rotMat3 = quaternionToRotMatrix(q_view)
    rotMat = torch.eye(4)[None, ...].repeat(
        [cfg.inputViewN, 1, 1]).to(cfg.device)
    rotMat[:, :3, :3] = rotMat3
    rotMat = rotMat.float().to(cfg.device)

    # 2D to 3D coordinate transformation
    khomMat = cfg.Khom2Dto3D # [4x4]
    khomMat = khomMat.repeat(

```

```

        [cfg.inputViewN, 1, 1]) # [B,V,4x4]

rotation_angle_post = np.deg2rad(90) # orig
rotation_axis_post = [0, 0, 1]

rotationPostMat = np.repeat(rotation_matrix(rotation_angle_post,
rotation_axis_post)[None, ...], cfg.inputViewN,
axis=0)
rotationPostMat = torch.from_numpy(rotationPostMat).float().to(cfg.device)

inverseAllTransMat = torch.inverse(torch.matmul(
    khomMat, torch.matmul(transMat, torch.matmul(rotationPostMat,
rotMat))))

# transform depth stack
ML = maskLogit.clone().reshape([chunkSize, 1, -1]) # [B,1,VHW]
XYZhom = get3DhomCoord(XYZ, cfg, chunkSize) # [B,V,4,HW]
XYZid = torch.matmul(inverseAllTransMat, XYZhom) # [B,V,3,HW]
XYZid = XYZid[:, :, :3, :] # [B,V,3,HW]

# fuse point clouds
XYZid = XYZid.permute([0, 2, 1, 3]).reshape(
    [chunkSize, 3, -1]) # [B,3,VHW]

return XYZid, ML

def quaternionToRotMatrix(q):
    # q = [V, 4]
    qa, qb, qc, qd = torch.unbind(q, dim=1) # [V,]
    R = torch.stack(
        [torch.stack([1 - 2 * (qc**2 + qd**2),
            2 * (qb * qc - qa * qd),
            2 * (qa * qc + qb * qd)]),
        torch.stack([2 * (qb * qc + qa * qd),
            1 - 2 * (qb**2 + qd**2),
            2 * (qc * qd - qa * qb)]),
        torch.stack([2 * (qb * qd - qa * qc),
            2 * (qa * qb + qc * qd),
            1 - 2 * (qb**2 + qc**2)])]

```

```

).permute(2, 0, 1)
return R.to(q.device)

def transParamsToHomMatrix(q, t):
    """q = [V, 4], t = [V,3]"""
    N = q.size(0)
    R = quaternionToRotMatrix(q) # [V,3,3]
    Rt = torch.cat([R, t.unsqueeze(-1)], dim=2) # [V,3,4]
    hom_aug = torch.cat([torch.zeros([N, 1, 3]), torch.ones([N, 1, 1])],
                        dim=2).to(Rt.device)
    RtHom = torch.cat([Rt, hom_aug], dim=1) # [V,4,4]
    return RtHom

def get3DhomCoord(XYZ, cfg, chunkSize):
    ones = torch.ones([chunkSize, cfg.inputViewN, cfg.outh, cfg.outW]) \
        .to(XYZ.device)
    XYZhom = torch.cat([XYZ, ones], dim=1) \
        .reshape([chunkSize, 4, cfg.inputViewN, -1]) \
        .permute([0, 2, 1, 3])

    return XYZhom # [B,V,4,HW]

```