Introduction to HOG3D

- HOG3D: Histogram of Oriented Gradients in 3D
 - Extension of 2D HOG for volumetric data

3D Gradient Computation

- Computes gradients in x, y, and z directions
- Each gradient represents the rate of change of intensity in that direction
- Smoothing (e.g., Gaussian filter) is often applied first to reduce noise

smoothed_data = gaussian_filter(ct_data, sigma=1)
gx, gy, gz = np.gradient(smoothed_data)

Magnitude Calculation

- Magnitude represents the strength of the gradient
- Higher magnitude indicates stronger edges or intensity changes
- Crucial for weighting the importance of gradients in later steps

magnitude = np.sqrt(gx **2 + gy **2 + gz **2)

Orientation Calculation - Phi (Elevation)

- Phi represents the elevation angle in spherical coordinates
- Measures the angle between the gradient vector and the z-axis
- Range: [0, π]
- Captures the "up-down" orientation of the gradient

phi = np.arctan2(np.sqrt(gx**2 + gy**2), gz)

Orientation Calculation - Theta (Azimuth)

- Theta represents the azimuth angle in spherical coordinates
- Measures the angle between the x-axis and the projection of the gradient vector onto the xy-plane
- Range: [-π, π]
- Captures the "left-right" orientation of the gradient



Understanding Spherical Coordinates

• Phi (Elevation):

- 0° = pointing straight up (+z direction)
- 90° = horizontal
- 180° = pointing straight down (-z direction)

• Theta (Azimuth):

- 0° = pointing along +x axis
- 90° = pointing along +y axis
- 180° or -180° = pointing along -x axis
- -90° = pointing along -y axis



Cells in HOG3D

- Cells are small 3D regions of the input volume
- Typical size: 4x4x4 or 8x8x8 voxels
- Purpose: Capture local gradient information
- Smaller than the object of interest, allowing for some spatial invariance

cell_phi = phi[x+i:x+i+cell_size, y+j:y+j+cell_size, z+k:z+k+cell_size]
cell_theta = theta[x+i:x+i+cell_size, y+j:y+j+cell_size, z+k:z+k+cell_size]
cell_magnitude = magnitude[x+i:x+i+cell_size, y+j:y+j+cell_size, z+k:z+k+cell_size]

Cell Histograms

- Compute histograms of gradient orientations for each cell
- Separate histograms for phi and theta
- Weighted by gradient magnitude
- Captures the distribution of edge directions within the cell

hist_phi, _ = np.histogram(cell_phi.ravel(), bins=num_bins, range=(-np.pi, np.pi), weights=cell_magnitude.ravel()



Blocks in HOG3D

- Blocks are larger 3D regions composed of multiple cells
- Typical size: 2x2x2 or 3x3x3 cells
- Purpose: Provide a broader context and allow for normalization

block_hist[:, cell_index] = np.concatenate([hist_phi, hist_theta])

Block Formation

- Iterate over cells within a block
- Concatenate phi and theta histograms for each cell
- Results in a descriptor for the entire block

```
for i in range(0, block_size, cell_size):
    for j in range(0, block_size, cell_size):
        for k in range(0, block_size, cell_size):
            # ... compute cell histograms ...
            block_hist[:, cell_index] = np.concatenate([hist_phi, hist_theta])
            cell_index += 1
```

Block Normalization

- Normalize the combined histograms within each block
- Improves robustness to illumination and contrast variations
- Helps in comparing features across different parts of the volume

block_hist = block_hist.ravel()
block_hist /= np.linalg.norm(block_hist) + 1e-5

Parameters

- **Downsample Factor**: Reduces input data size (e.g., 4 means 1/4 size in each dimension)
- **Block Size**: Size of each 3D block (e.g., 4 means 4x4x4 voxels)
- **Cell Size**: Size of each cell within a block (e.g., 2 means 2x2x2 voxels)
- Stride: Step size between blocks (controls overlap)
- Num Bins: Number of orientation bins for histograms

block_size = 4
cell_size = 2
stride = 2
num_bins = 18

Final HOG3D Descriptor

- Concatenate normalized block histograms from across the volume
- Results in a high-dimensional feature vector
- Captures both local and semi-local 3D gradient information

hog3d_features.append(block_hist)
hog3d_positions.append([x + block_size // 2, y + block_size // 2, z + block_size // 2])
hog3d_orientations.append([np.mean(cell_phi), np.mean(cell_theta)])

Breakdown of Features

Breakdown of 288 Features

 The number 288 in the HOG3D feature vector represents the total number of features per block.

Here's how it's calculated:

- Number of orientation bins: 18 for phi + 18 for theta = 36 bins total
- Number of cells in a block: (4 // 2)^3 = 2^3 = 8
 cells
- Cell Phi Bins Theta Bins Cell 1 Features 1-18 Features 19-36 Features 55-72 Cell 2 Features 37-54 Cell 3 Features 73-90 Features 91-108 Cell 4 Features 109-126 Features 127-144 Cell 5 Features 145-162 Features 163-180 Cell 6 Features 181-198 Features 199-216 Cell 7 Features 217-234 Features 235-252 Cell 8 Features 253-270 Features 271-288

• Features per block: 36 * 8 = 288

Visualizing Features with respect to Positions

