

Answer:

(a) Steps for Obtaining Optimized Points-Based Motion Parallax with Full 3D Motion:

(a) 通过全 3D 运动获得优化的基于点的运动视差的步骤:

To obtain optimized motion parallax in a scenario where the camera undergoes full 3D motion (both translation and rotation), the following steps are undertaken:

为了在相机进行全 3D 运动（平移和旋转）的场景中获得优化的运动视差，需要执行以下步骤:

1. Image Acquisition:图像采集:

- **Capture Images:** Collect a sequence of images as the camera moves through the environment, experiencing both translation and rotation. This could be from a moving vehicle, a handheld camera, or any moving platform.  
**捕捉图像:** 当相机在环境中移动、经历平移和旋转时收集一系列图像。这可能来自移动的车辆、手持相机或任何移动平台。
- **Camera Calibration:** Ensure the camera is calibrated to know its intrinsic parameters (focal length, principal point, distortion coefficients).  
**相机标定:** 确保相机经过标定，了解其内在参数（焦距、主点、畸变系数）。

2. Preprocessing:预处理:

- **Undistort Images:** Correct for lens distortion using the camera's intrinsic parameters to ensure accurate feature detection and matching.  
**图像不失真:** 使用相机的内在参数校正镜头失真，以确保准确的特征检测和匹配。

3. Feature Detection and Description:特征检测及描述:

- **Detect Keypoints:** Use feature detectors such as SIFT, SURF, ORB, or Harris corner detectors to find salient points in each image.  
**检测关键点:** 使用 SIFT、SURF、ORB 或 Harris 角点检测器等特征检测器来查找每个图像中的显著点。
- **Compute Descriptors:** For each keypoint, compute a descriptor that uniquely characterizes the local image region around the keypoint.  
**计算描述符:** 对于每个关键点，计算一个描述符，该描述符唯一地表征关键点周围的局部图像区域。

4. Feature Matching:特征匹配:

- **Match Features Across Images:** Use descriptor matching techniques (e.g., nearest-neighbor search, FLANN matcher) to find correspondences between keypoints in different images.  
**跨图像匹配特征:** 使用描述符匹配技术（例如，最近邻搜索、FLANN 匹配器）来查找不同图像中关键点之间的对应关系。
- **Outlier Rejection:** Apply methods like Lowe's ratio test and cross-validation to eliminate false matches.  
**异常值拒绝:** 应用 Lowe 比率测试和交叉验证等方法来消除错误匹配。

5. Initial Motion Estimation:

- **Compute the Essential or Fundamental Matrix:计算基本矩阵:**
  - For **calibrated cameras**, compute the **Essential matrix** using the matched points.对于**校准相机**，使用匹配点计算基本矩阵。
  - For **uncalibrated cameras**, compute the **Fundamental matrix**.对于**未校准的相机**，计算基本矩阵。
- **Decompose the Essential Matrix:分解基本矩阵:**
  - Extract the relative **rotation (R)** and **translation (t)** between the camera positions.提取相机位置之间的相对**旋转 (R)**和**平移 (t)**。
  - This involves singular value decomposition (SVD) of the Essential matrix.这涉及基本矩阵的奇异值分解（SVD）。

6. Triangulation:三角测量:

- **Reconstruct 3D Points:** Use the relative camera poses (R and t) and the matched image points to triangulate the 3D positions of the feature points.  
**重建 3D 点:** 使用相对相机位姿（R 和 t）和匹配的图像点对特征点的 3D 位置进行三角测量。
- **Linear Triangulation:** Apply methods like the Direct Linear Transformation (DLT) to compute the 3D coordinates.  
**线性三角测量:** 应用直接线性变换 (DLT) 等方法来计算 3D 坐标。

7. Optimization (Bundle Adjustment):优化（捆绑调整）:

- **Set Up the Optimization Problem:设置优化问题:**
  - Define a cost function based on the **reprojection error**, which is the difference between the observed image points and the projections of the estimated 3D points.  
**基于重投影误差定义成本函数**，该误差是观察到的图像点与估计 3D 点的投影之间的差异。
- **Simultaneous Refinement:同时细化:**
  - Adjust the camera parameters (**rotations and translations**) and the **3D point positions** to minimize the reprojection error.  
**调整相机参数（旋转和平移）和 3D 点位置**以最小化重投影误差。
- **Use Non-Linear Optimization:使用非线性优化:**
  - Apply algorithms like Levenberg-Marquardt for optimization.  
**应用 Levenberg-Marquardt 等算法进行优化。**
- **Incorporate All Views:合并所有视图:**
  - Perform a global optimization over all images and 3D points to ensure consistency.对所有图像和 3D 点执行全局优化以确保一致性。

8. Outlier Detection and Removal:异常值检测和去除:

- **RANSAC Algorithm:搜索算法:**
  - Use RANSAC to robustly estimate the motion parameters by iteratively fitting the model and removing points that do not fit well.  
**使用 RANSAC 通过迭代拟合模型并删除不太拟合的点来鲁棒地估计运动参数。**
- **Refine Matches:细化匹配:**
  - Exclude outliers from the dataset to improve the accuracy of the reconstruction.  
**从数据集中排除异常值以提高重建的准确性。**

9. Scale Estimation (if Necessary):规模估计（如果需要）:

- **Monocular Scale Ambiguity:**
  - In monocular setups, the reconstruction is up to an unknown scale. Introduce known measurements or constraints to resolve the scale ambiguity.  
**在单眼设置中，重建达到未知的规模。引入已知的测量或约束来解决尺度模糊性。**

10. Motion Parallax Analysis:运动视差分析:

- **Compute Parallax Vectors:**
  - Analyze the apparent displacement of 3D points between frames due to the camera's motion.分析由于相机运动而导致的帧间 3D 点的明显位移。
- **Depth Estimation:**
  - Use the parallax information to infer the relative depths of objects in the scene.  
**使用视差信息来推断场景中对象的相对深度。**

11. Visualization and Interpretation:可视化和解释:

- **Generate Depth Maps or Point Clouds:生成深度图或点云:**
  - Visualize the reconstructed 3D structure to interpret the spatial arrangement of the scene.可视化重建的 3D 结构以解释场景的空间排列。
- **Interpret Motion Parallax:**
  - Understand how objects at different depths move relative to each other due to the ego-motion.了解不同深度的物体如何由于自我运动而相对移动。

12. Iterative Refinement:迭代细化:

- **Loop Back if Necessary:如果需要则循环返回:**
  - If the optimization does not converge satisfactorily, revisit previous steps with adjusted parameters or improved feature detection.  
**如果优化未令人满意地收敛，请使用调整的参数或改进的特征检测重新访问先前的步骤。**

Summary:概括:

By following these steps, one can obtain an optimized estimation of both the camera's motion and the 3D structure of the scene, accounting for the full 3D motion (translation and rotation). The optimization process ensures that the motion parallax is accurately represented, allowing for precise depth perception and scene reconstruction.

通过遵循这些步骤，人们可以获得相机运动和场景 3D 结构的优化估计，并考虑完整的 3D 运动（平移和旋转）。优化过程可确保准确表示运动视差，从而实现精确的深度感知和场景重建。

(b) Derivation of the Full 3D Location of a Point by Image Disparity from a Parallel Binocular Image Pair:(b) 根据平行双目图像对的图像视差推导点的完整 3D 位置:

In a parallel binocular stereo setup, two cameras with parallel optical axes and aligned image planes capture images of the same scene from slightly different positions. The key parameters are:

在平行双目立体设置中，具有平行光轴和对齐图像平面的两个相机从略有不同的位置捕获同一场景的图像。关键参数是:

- **Baseline (B):** The distance between the optical centers of the two cameras.  
**基线 (B) :** 两个相机的光学中心之间的距离。
- **Focal Length (f):** The focal length of the cameras (assumed identical).  
**焦距 (f):** 相机的焦距（假设相同）。
- **Disparity (d):** The difference in the x-coordinate of the corresponding points in the left and right images.**视差 (d) :** 左右图像中对应点的x坐标之差。

Coordinate System Setup:

- The left camera is positioned at the origin (0, 0, 0).左相机位于原点 (0, 0, 0) 。
- The right camera is positioned at (B, 0, 0).右侧摄像头位于 (B, 0, 0) 。
- The 3D point P has coordinates (X, Y, Z).3D点 P 有坐标 (X, Y, Z) 。

Projection Equations:投影方程:

- For the **left camera**:对于**左侧摄像头**:
$$x_L = \frac{fX}{Z}, \quad y_L = \frac{fY}{Z}$$
- For the **right camera**:对于**正确的相机**:
$$x_R = \frac{f(X - B)}{Z}, \quad y_R = \frac{fY}{Z}$$

**Note:** Since the cameras are aligned,  $y_L = y_R$ .**注意:** 由于相机已对齐， $y_L = y_R$  。

Deriving Disparity (d):

- Disparity is defined as:视差定义为:
$$d = x_L - x_R$$
- Substituting the projection equations:代入投影方程:
$$d = \frac{fX}{Z} - \frac{f(X - B)}{Z} = \frac{fX}{Z} - \left( \frac{fX}{Z} - \frac{fB}{Z} \right) = \frac{fB}{Z}$$

Solving for Depth (Z):

- Rearranging the disparity equation:重新整理视差方程:
$$Z = \frac{fB}{d}$$

Computing the Full 3D Coordinates:计算完整 3D 坐标:

- **X-coordinate:**
$$X = \frac{x_L Z}{f} = \frac{x_L}{f} \cdot \frac{fB}{d} = \frac{x_L B}{d}$$
- **Y-coordinate:**
$$Y = \frac{y_L Z}{f} = \frac{y_L}{f} \cdot \frac{fB}{d} = \frac{y_L B}{d}$$
- **Z-coordinate:**
$$Z = \frac{fB}{d}$$

Summary Equation:总结方程:

The full 3D coordinates of point P are:点的完整 3D 坐标 P 是:

$$\begin{cases} X = \frac{x_L B}{d} \\ Y = \frac{y_L B}{d} \\ Z = \frac{fB}{d} \end{cases}$$

Interpretation:

- **Depth (Z):** Inversely proportional to the disparity d. Larger disparities correspond to closer objects.**深度 (Z) :** 与视差成反比 d 。较大的差异对应于较近的物体。

- **Horizontal Position (X):** Directly proportional to the left image's x-coordinate and inversely proportional to disparity.**水平位置 (X):** 与左图像的 x 坐标成正比，与视差成反比。

- **Vertical Position (Y):** Directly proportional to the left image's y-coordinate and inversely proportional to disparity.**垂直位置 (Y):** 与左图像的 y 坐标成正比，与视差成反比。

Conclusion:

By measuring the disparity between corresponding points in the left and right images of a parallel binocular stereo pair and knowing the cameras' focal length and baseline, the full 3D position of points in the scene can be accurately computed. This process effectively reduces the general motion parallax problem to a stereo vision problem, leveraging the fixed geometry of the stereo setup to simplify calculations.

通过测量平行双目立体对的左右图像中对应点之间的视差并了解相机的焦距和基线，可以准确计算场景中点的完整 3D 位置。该过程有效地将一般运动视差问题简化为立体视觉问题，利用立体设置的固定几何形状来简化计算。