

Answers

Part 1: System Design

Q1. Max exposure time for < 2 px blur at 100 MPH

- Ball Velocity $v = 100 \text{ MPH} = 44.704 \text{ m/s}$
- Standard baseball diameter $D = 9 \text{ in} / \pi = 2.86 \text{ in} = 0.0726 \text{ m}$
- Assuming conditions from Q2, ball occupies > 200 px:

Spatial Resolution (R) = 200 px / 0.0726 m = 2754.8 px/m

Image Velocity (v_{px}) = $v * R = 44.704 * 2754.8 = 123,151 \text{ px/s}$

We want max blur = 2 pixels. Thus:

$$t_{exp} <= 2/v_{px} = 2/123151 = 1.624 \times 10^{-5} \text{ s}$$

$$t_{exp} <= 16.24 \mu\text{s} \text{ (or about } 1/61,500 \text{ shutter speed)}$$

Q2. Sensor resolution & focal length (5m away, 200px diameter)

- Distance $Z = 5 \text{ m}$; Diameter $D = 0.0726 \text{ m}$; Image $d = 200 \text{ px}$
- Using pinhole camera relation: $d / f_{px} = D / Z$

Assignment Answers (Cont.)

Q3. Hough Circle Transform vs YOLO (Deep Learning)

- Hough Circle is deterministic, runs fast on CPUs, and is geometry-based.

* Pro: Computationally cheap, real-time edge device friendly.

* Con: Fails easily on motion blur, shadows, or if ball shape distorts.

- YOLO/Seg is purely data-driven, learning semantic representations.

* Pro: Highly robust to blur, varied lighting, and background noise.

* Con: Requires GPU/NPU for real-time (< 5ms) processing, and

requires gathering/labeling thousands of baseball frames.

Q4. Visual challenge of 'Bullet Spin' (Axis parallel to Camera)

- Challenge: The 3D seam pattern doesn't 'rotate around' the visible

hemisphere of the ball. It only rotates 2D in the image plane.

- Because depth cues (seams vanishing/appearing over the horizon)

are absent, estimating the true 3D spatial phase is ill-posed.

- Logic Change: Instead of mapping seam translation across a 3D sphere,