

# 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:

$$\text{Spatial Resolution (R)} = 200 \text{ px} / 0.0726 \text{ m} = 2754.8 \text{ px/m}$$

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

We want max blur = 2 pixels. Thus:

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

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

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:  $d / f_{\text{px}} = D / Z$

## 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 ( $< 5\text{ms}$ ) 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,