

Addressing Challenges with Augmented Reality Applications on Smartphones

Krzysztof Zienkiewicz

krzysztof.k.zienkiewicz@vanderbilt.edu

**Vanderbilt University
Nashville, Tennessee**



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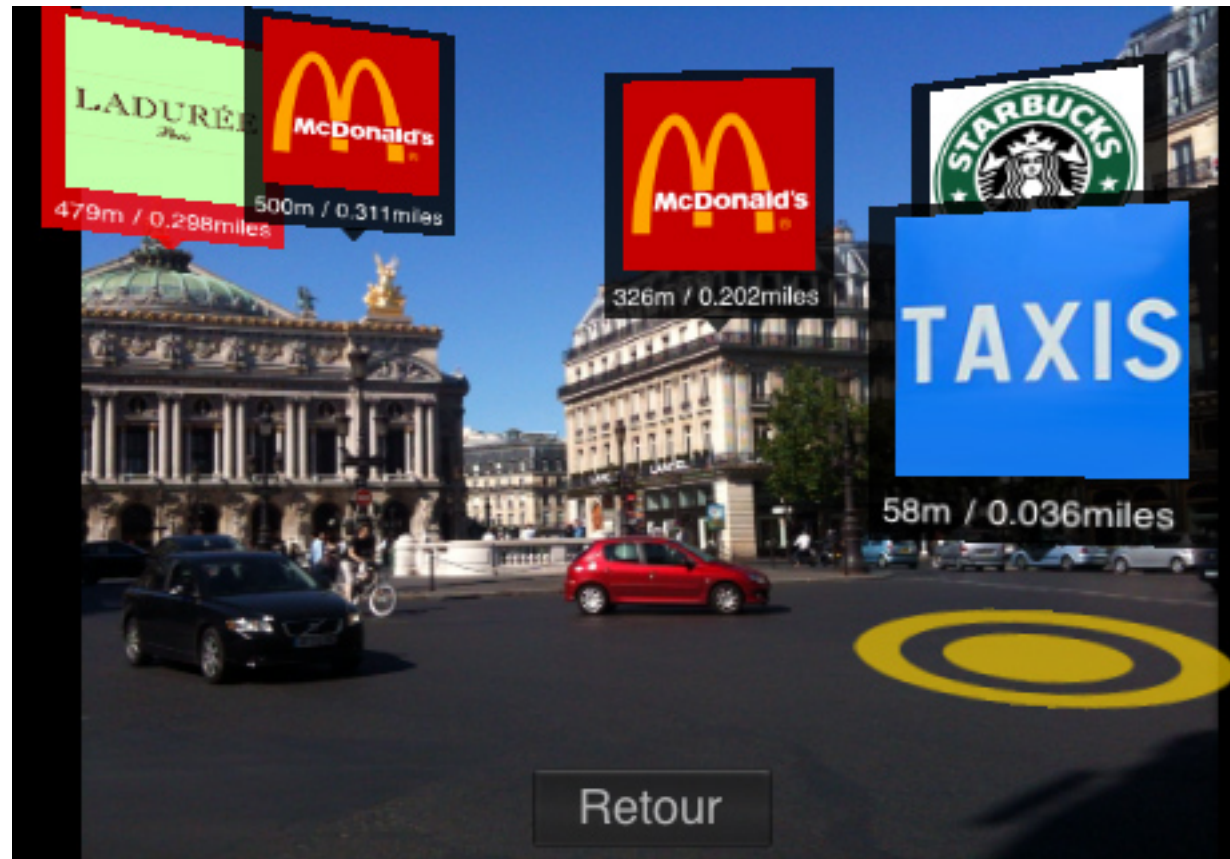
Augmented Reality (AR)

- Azuma's definition: AR is a system that has the following characteristics
 - Combines real and virtual
 - Interactive in real time
 - Registered in 3D
- Milgram's Continuum



Components of an AR application

- Rendering 3D context-specific data
- Retrieving geotagged Points Of Interest (POIs)
- Frame of Reference estimation



Challenges

- Mobile 3D solutions are not optimal and hard to mesh with camera imagery
- Filtering geotagged POIs by proximity is computationally intensive
- Real-time estimation of frame of reference is computationally demanding
- Geomagnetic sensor noise makes orientation estimation hard

Mobile 3D solutions are not optimal

- AR needs precise 3D overlays to provide an intuitive interface
- Without accuracy, user experience deteriorates
- Meshing 3D data with camera preview is highly platform dependent
- Mobile GLs don't offer a "pick" mode

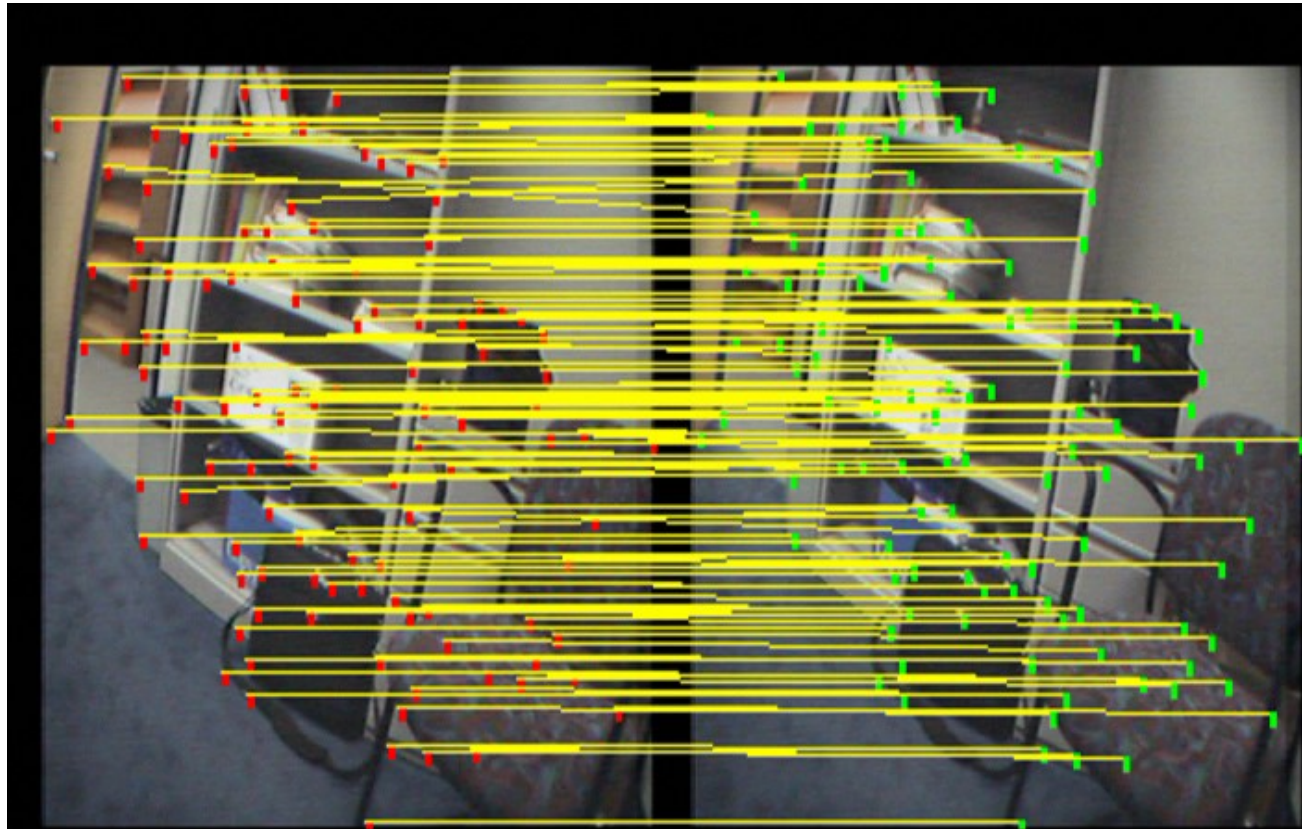


Retrieving POIs by proximity is expensive

- Areas of interest may have a high density of POIs
- Retrieval by location requires many comparisons
- Bandwidth limitations

Estimation of frame of reference

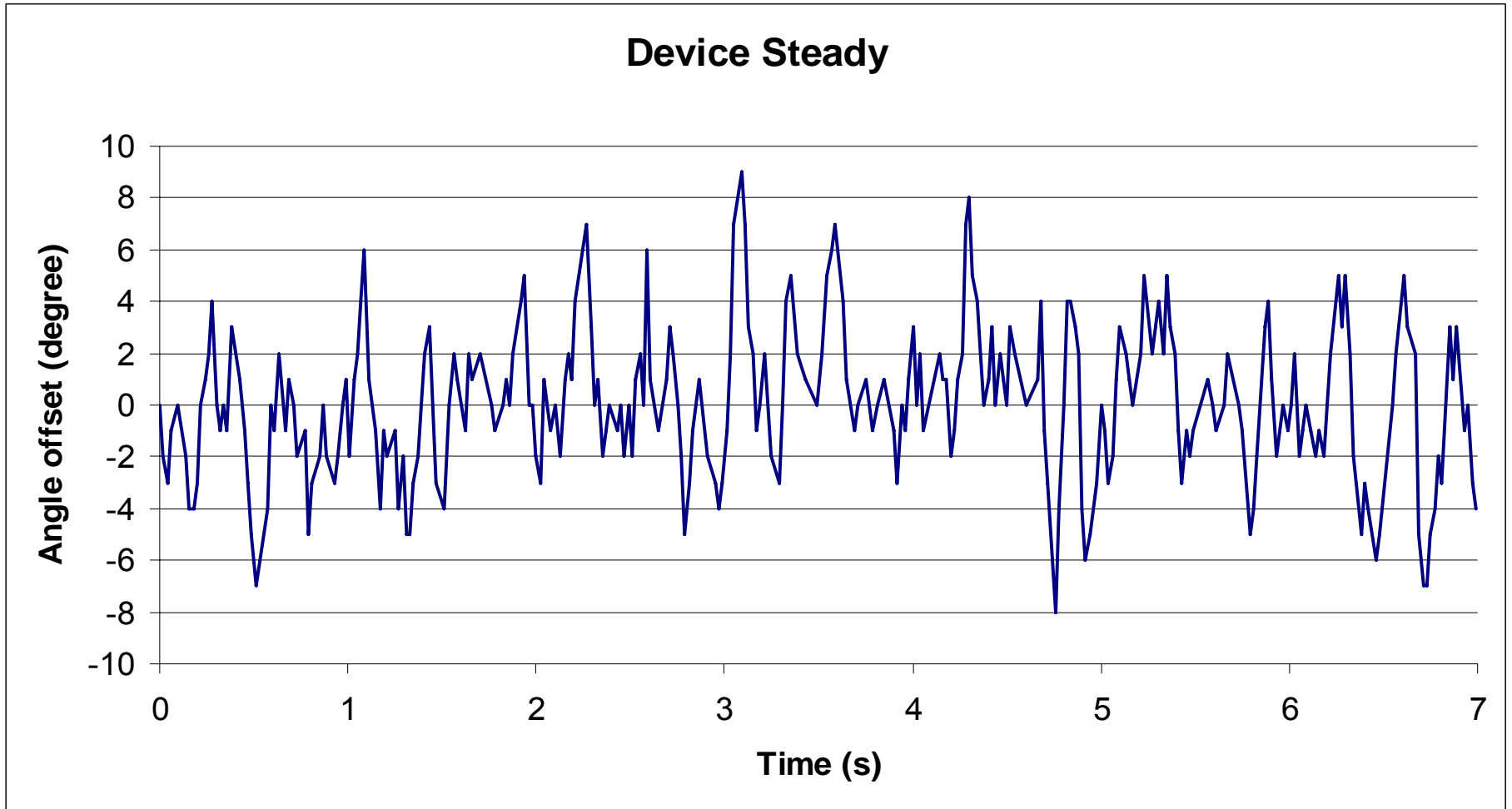
- Fiducial markers
- Feature extraction



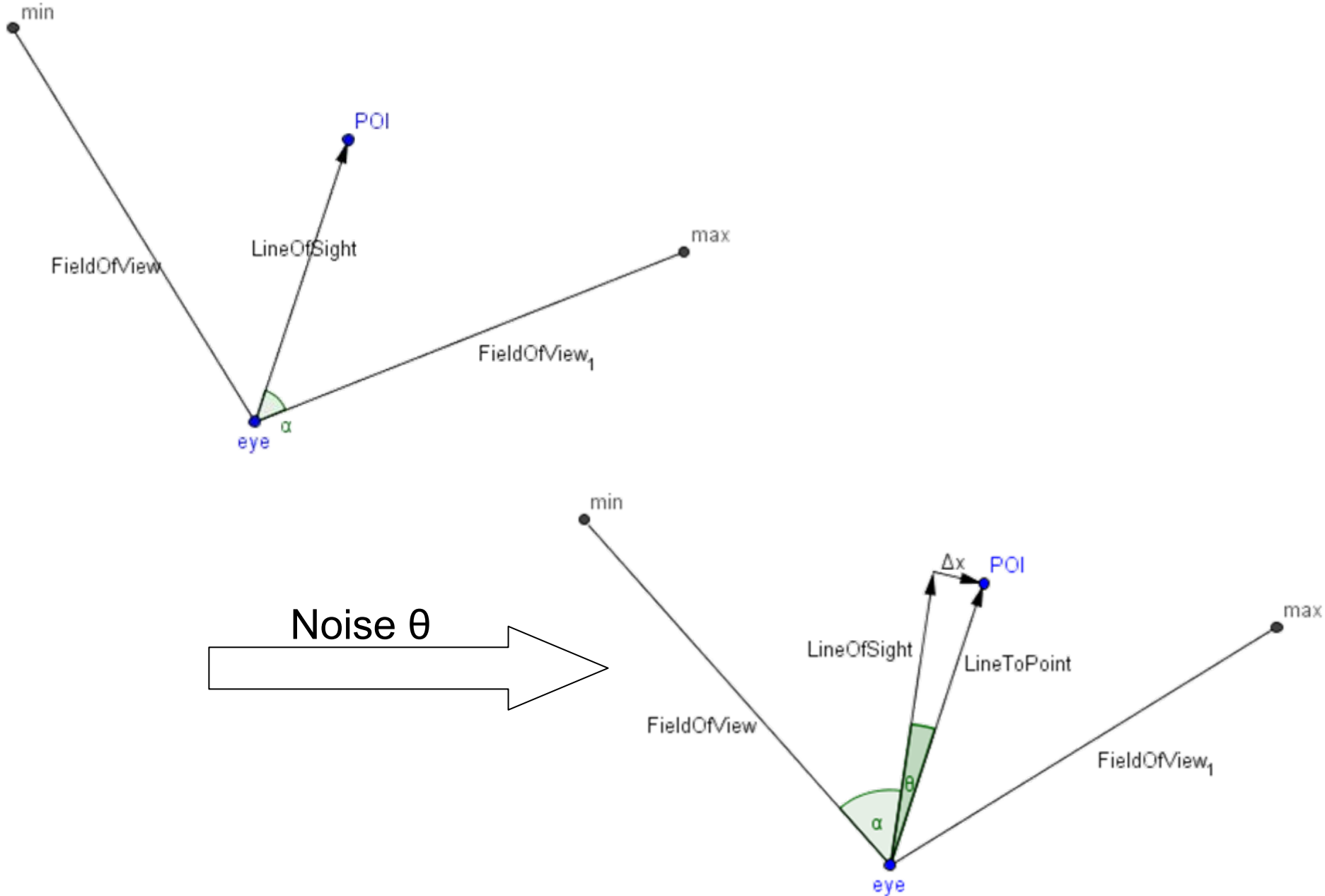
Geomagnetic sensor noise

- Compass used to estimate the field of vision
- Compass noise introduces jitter in the rendered overlays
- Savitzky-Golay smoothing filter is too slow
- Noise +/- 9 degrees

Geomagnetic sensor noise

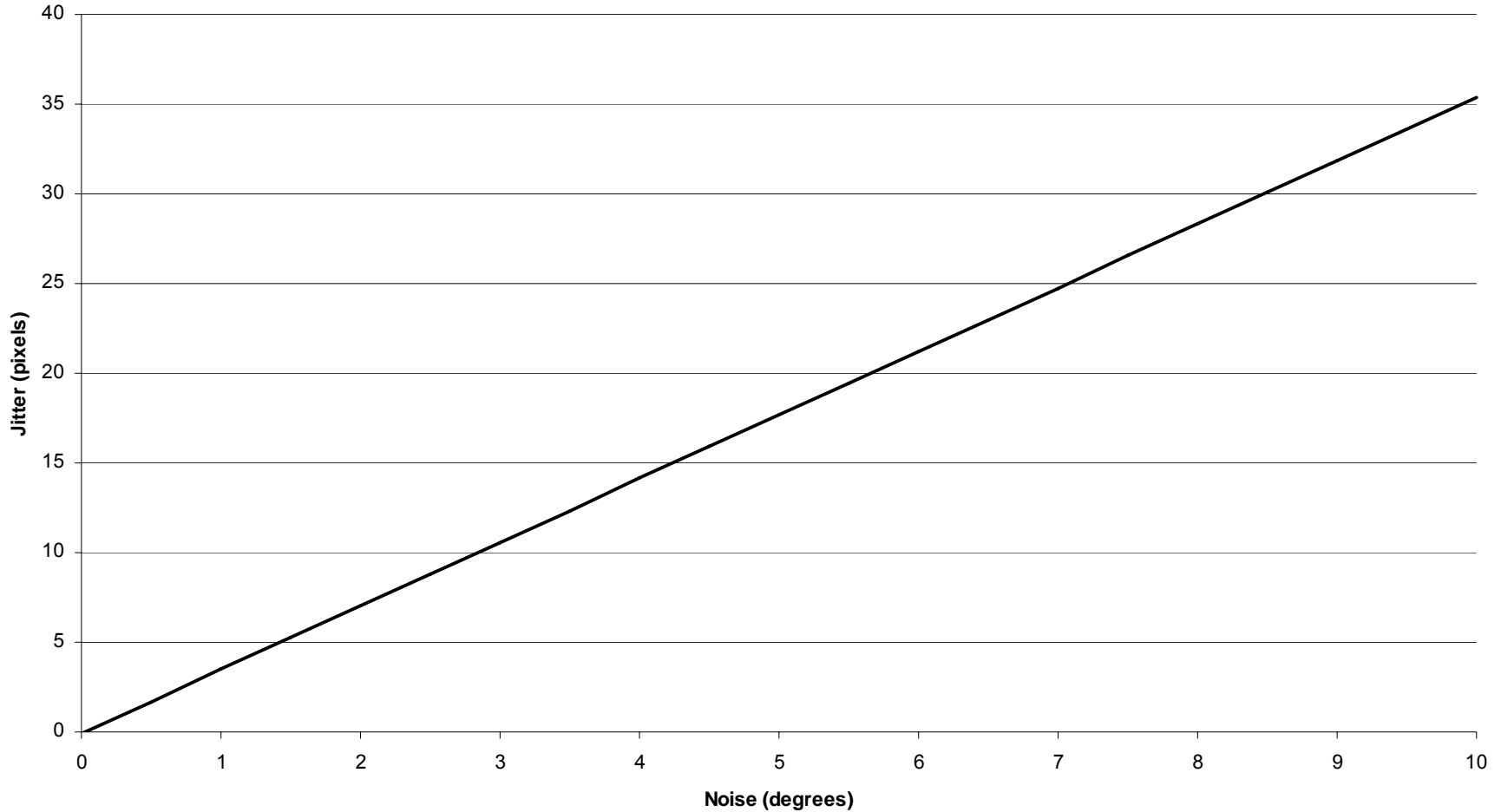


Effects of geomagnetic sensor noise



Effects of geomagnetic sensor noise

Effect of Compass Noise



$$jitter = \frac{width}{2 \tan \alpha} \tan \theta$$

Solutions

- Using hardware accelerated 3D APIs to display rendered content
- A grid-based approach to data storage and retrieval
- Using GPS and geomagnetic sensors to estimate device position and orientation
- Statistical analysis filter

Hardware accelerated 3D APIs

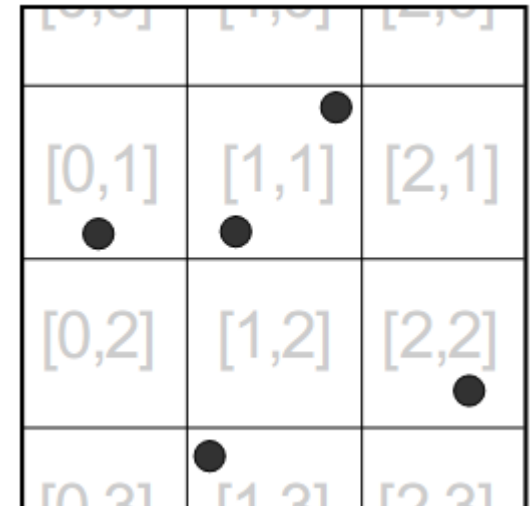
- Apple's UIKit allows a transformation matrix to be applied to views
 - Allows for hit testing
- Other platforms will need to do hit testing by hand
 - But get to use GLs for rendering
- Meshing achieved via layering

$$\begin{bmatrix} \frac{F}{\text{aspect}} & 0 & 0 & 0 \\ 0 & F & 0 & 0 \\ 0 & 0 & \frac{n+f}{n-f} & \frac{2fn}{n-f} \\ 0 & 0 & -1 & 0 \end{bmatrix}$$

Grid-based data storage and retrieval

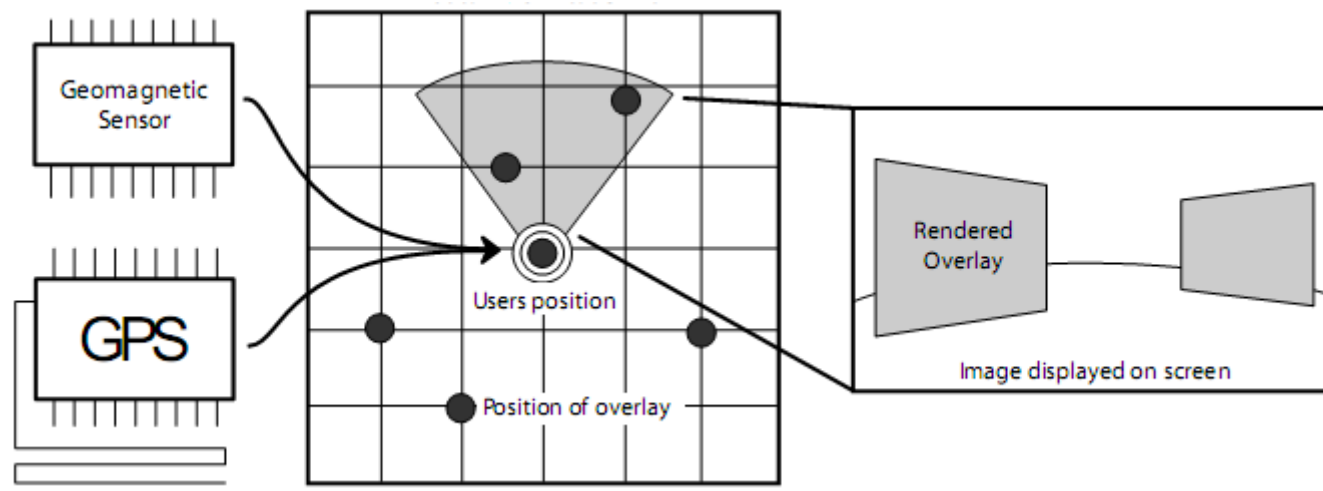
- Mapping function takes location to grid indices
- Each square contains all POIs within the region
- No numeric comparisons are performed by the server
- POIs are downloaded in bulk and need to be filtered on the phone

Query Type	Response Time
Latitude range (latitude column indexed)	581700 μ s
Latitude range (latitude column not indexed)	284600 μ s
Specific latitude bucket	5209 μ s



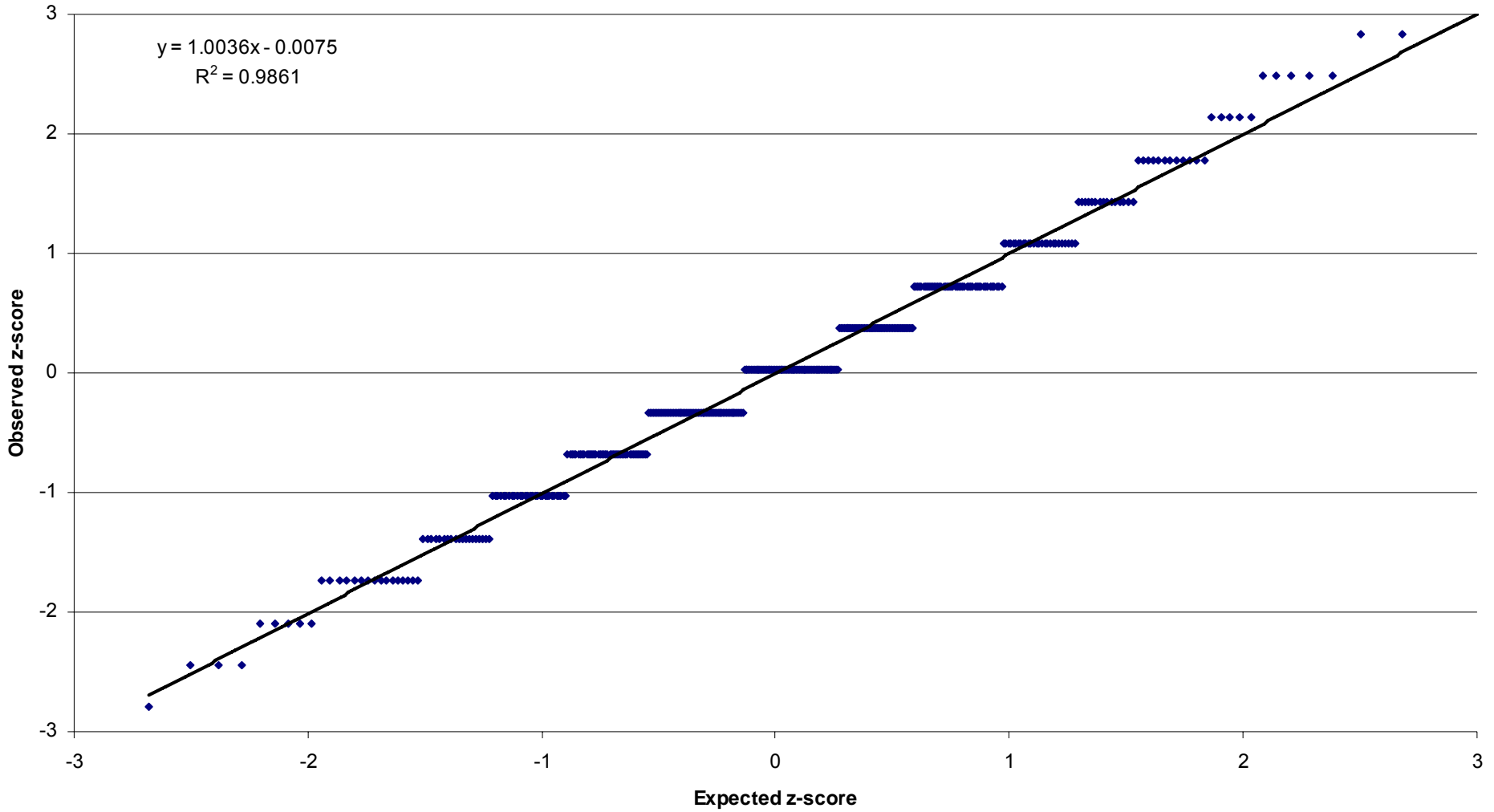
Using sensors to estimate field of view

- The computationally cheap solution
- Only need to update the transformation matrix



Nature of the noise

Noise Normality Test



Filtering algorithm

Variables/Functions:

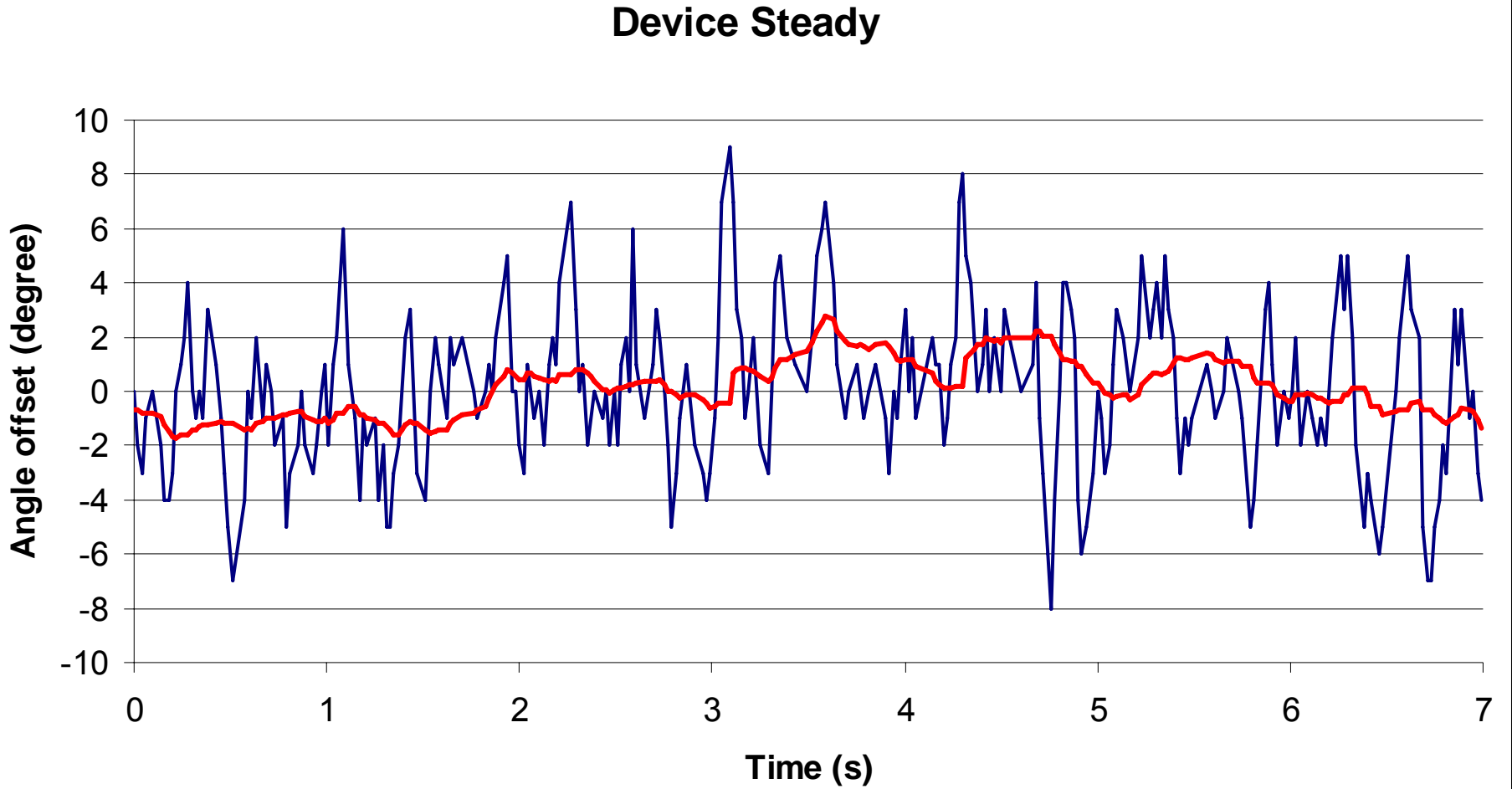
R = Ring Buffer of Received Data
 O = Ring Buffer of Outlier Data
 $|R| = |O|$ = Maximum Allowable Size of Buffer
 $size(buffer)$ = ReturnsCurrentSizeofBuffer
 p_i = A compass reading as a Single Precision Float
 $Z(p_i) = (p_i - mean(R)) / stdDev(R)$
 Z_{range} = Maximum Allowable Deviation
 $outlierDirection(p_i) = p_i > mean(R) ? 1 : -1$
 $enqueue(buffer, p_i)$ = Adds p_i to the Buffer

Algorithm:

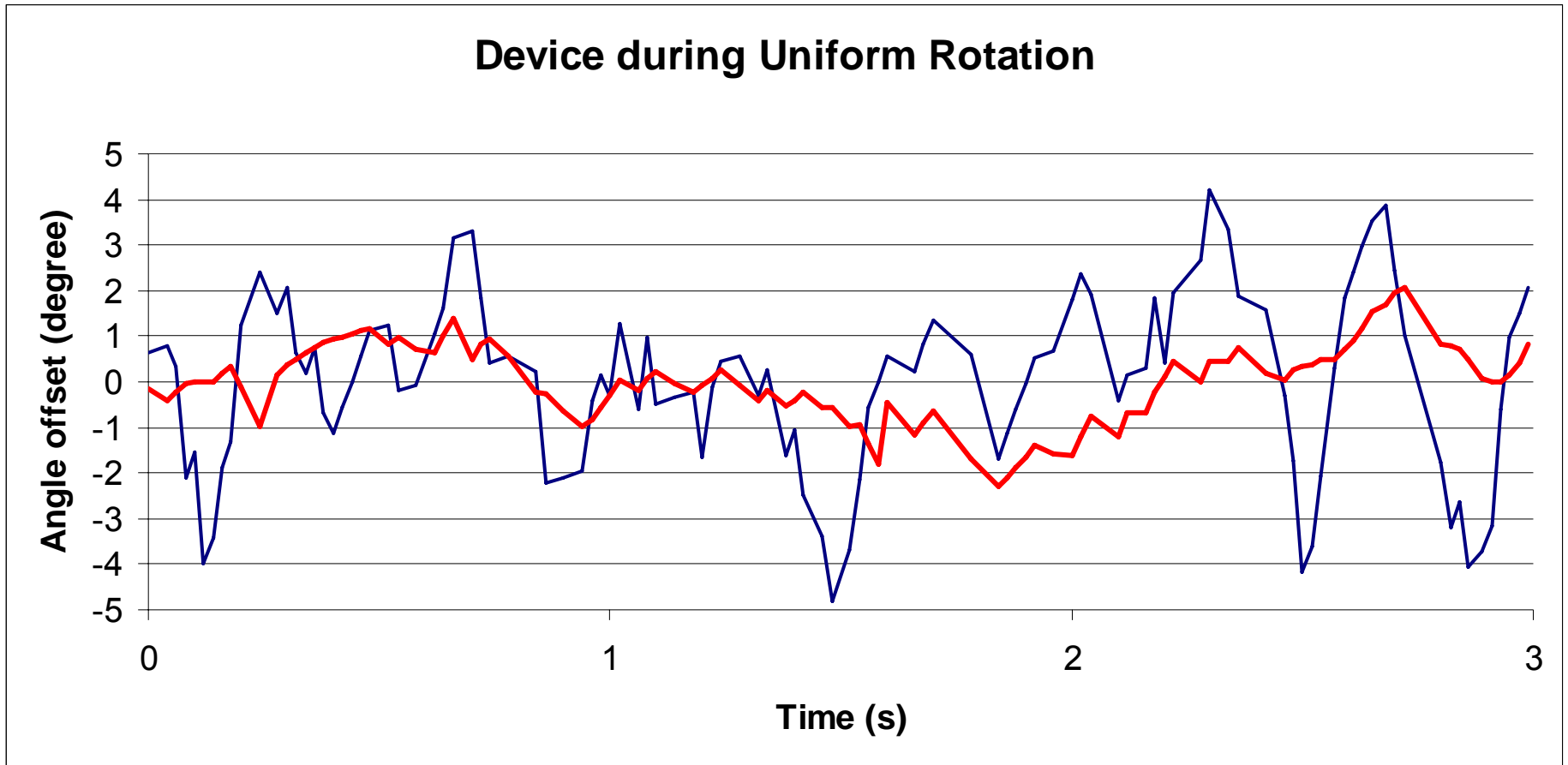
```
filtered( $p_i$ ) =  
  if  $size(R) < |R|$ : enqueue( $R, p_i$ )  
  else:  
     $z_i = Z(p_i)$   
    if  $abs(z_i) \leq Z_{range}$ :  
      enqueue( $R, p_i$ )  
      clear( $O$ )  
    else: enqueue( $O, p_i$ )  
  if  $size(O) = |O|$ :  
    side = outlierCluster()  
     $\forall p_j \in O$   
      if  $outlierDirection(p_j) = side$ :  
        enqueue( $R, p_j$ )  
        clear( $O$ )  
  return mean( $R$ )  
outlierCluster() =  
  int sum = 0  
   $\forall p_j \in O$   
     $sum += p_j - mean(R)$   
  return signum(sum)
```

- Small number of parameters
- Extendable for dynamic parameterization
- Good results: 60% noise reduction

Filtering algorithm



Filtering algorithm



Thanks

Questions?