

# Analysis of Head Pose as an Indicator of Driver's Visual Attention

Sumit Jha and Carlos Busso

Multimodal Signal Processing (MSP) Laboratory  
Department of Electrical Engineering,  
The University of Texas at Dallas,  
Richardson TX-75080, USA  
sxj146830@utdallas.edu, busso@utdallas.edu



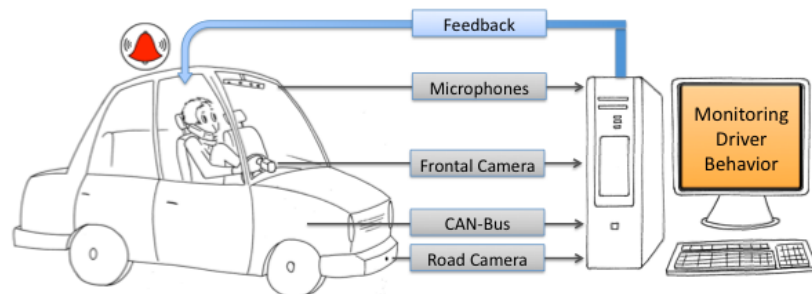


# Detecting Driving Distraction at MSP

- Key Problems:
  - Detect inattentive drivers using noninvasive sensors
  - Study realistic scenarios with real car driving in real roads

Nanxiang Li, Jinesh J. Jain, and Carlos Busso, "**Modeling of driver behavior** in real world scenarios using multiple noninvasive sensors," IEEE Transactions on Multimedia, vol. 15, no. 5, pp. 1213-1225, August 2013.

Nanxiang Li and Carlos Busso, "**Predicting perceived visual and cognitive distractions** of drivers with multimodal features," IEEE Transactions on Intelligent Transportation Systems, vol. 16, no. 1, pp. 51-65, February 2015.





# Drivers' Visual Attention

- Primary driving related task
  - Mirror checking actions (situational awareness)
  - Lane change
  - Turns and cross sections
- Secondary tasks
  - Visual Distraction for longer duration or one with high angle generally involves more head movement.[Zhang, 2008]
  - Cognitive distraction (“*looking but not seeing*”)



Nanxiang Li and Carlos Busso, "**Detecting drivers' mirror-checking actions** and its application to maneuver and secondary task recognition," IEEE Transactions on Intelligent Transportation Systems, vol. To appear.



# Motivations



- Gaze detection is a challenging problem in car environment
- It is often approximate by head position [Lee et al., 2011]
- Coarse direction of driver's gaze is enough for most in-vehicle applications [Tawari & Trivedi, 2014; Doshi & Trivedi, 2009]



Left mirror



Right mirror



Rear mirror

- Goal of this study is to analyze the relationship between gaze and head pose





# Objective

- Questions

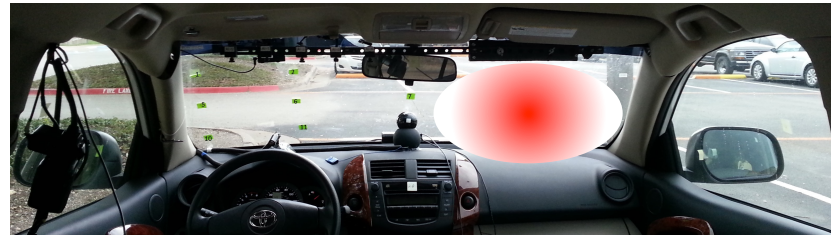


- How well does the head pose of the driver represent his/her gaze (visual attention)?



- How much does the head pose varies when the driver is looking at a certain direction?

- Can we define a confidence map for the drivers' gaze (visual attention) using head pose?





# Experimental Design

- Markers placed at the windshield and other locations of visual interest (mirrors, blind spot) in the UTDrive platform
- Participants are asked to look at markers

**Unconstrained:** “look at the point #4”

- When the car is parked
- When the subject is driving

**Constrained:** “head pose directed towards the mark”

- When the car is parked





# Unconstrained Condition

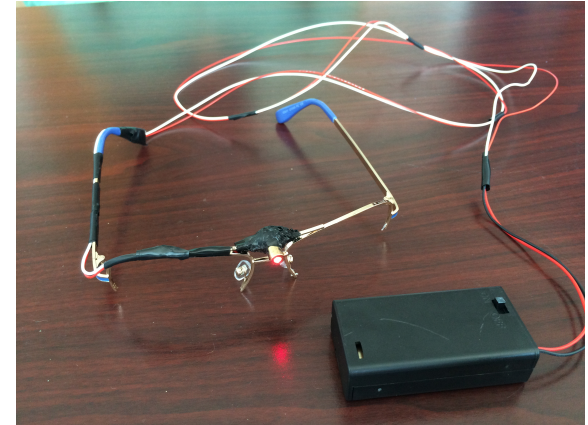
- Natural head poses
- Driver asked to look at various locations without further instructions
  - Collected when the car is parked and when driving
  - Multiple reading for each mark to capture the variance





# Constrained Collection

- We use a glasses' frame with laser pointer at the center
- The participants wore the frame, pointing to the target marks
  - Helps to establish a reference head pose without bias due to pupil movement (i.e., head pose = gaze)





# Data Collection

- Cameras placed to record the driver's face and the road scene.
- Used commercially available Dash camera (Blackvue dr650gw 2 channel) for the recording
  - Front camera for the road and rear camera for driver's face
  - Also records GPS and accelerometer data



- Pilot recording with four participants
  - Markers are randomly asked
  - Operator marked when the subject looks at markers





# Data Collection (cont.)



Unconstrained  
parked



Unconstrained  
driving



Constrained  
parked



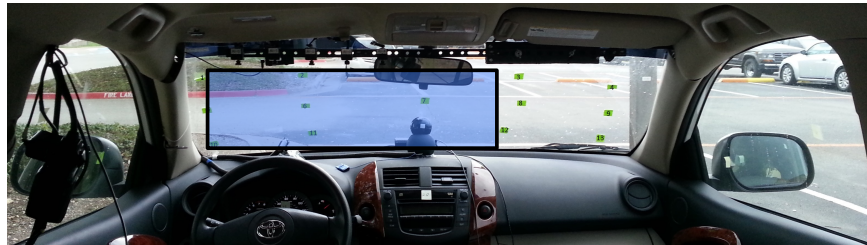


# Head Pose Estimation

- Used CMU's Intraface tool for head pose estimation
  - Yaw, pitch and roll



Very accurate when head pose is directed toward this area



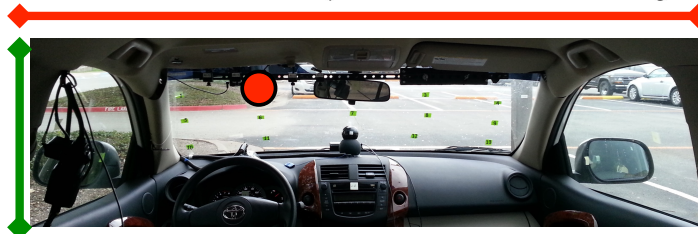
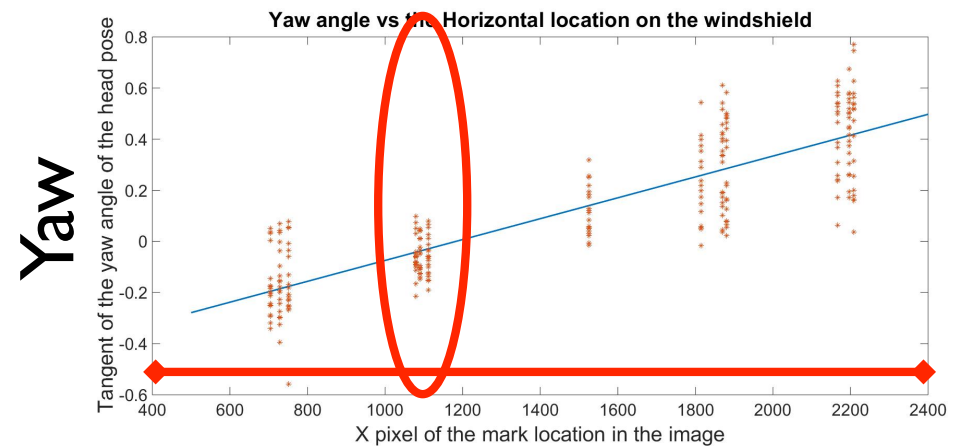
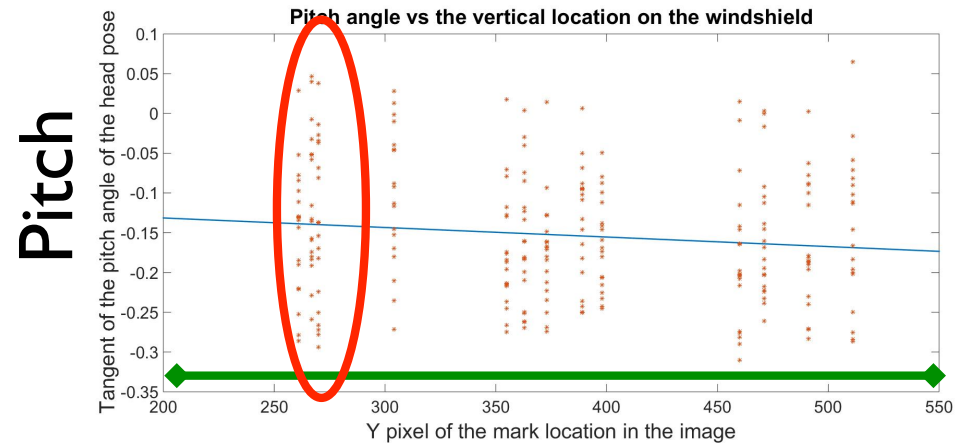


# Head pose variability per marker



- Observations

- Pitch highly unpredictable, low correlation with gaze direction
- Yaw angle has high correlation
  - The mean of the angles have linear relationship
  - The variance increases as the gazes direction increases





# Mapping head pose to markers

- We train regression model to map head pose into x,y coordinates
  - Input variable: head pose angle
  - Output variable: x or y coordinate in the windshield (in pixels)

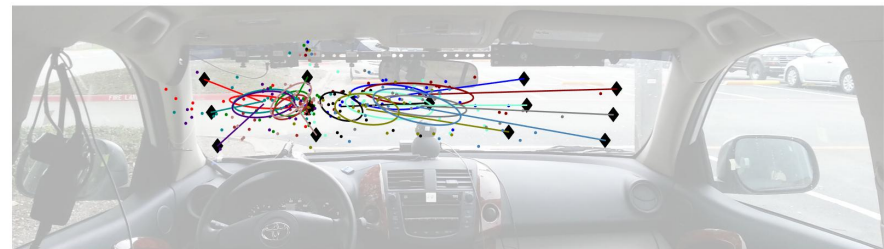
$$\textit{Estimated}_x = \alpha_0 + \alpha_1 \tan(\textit{yaw})$$

$$\textit{Estimated}_y = \alpha_0 + \alpha_1 \tan(\textit{pitch})$$

- We use constrained recordings where we know the intended coordinates
  - Ellipsoids define confidence regions



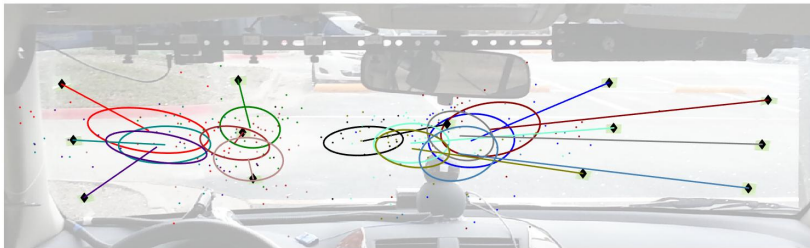
Unconstrained - parked



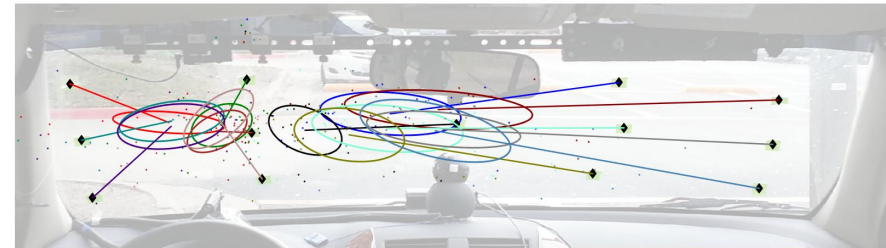
Unconstrained - driving



# Mapping head pose to markers (cont.)



Unconstrained - parked



Unconstrained - driving

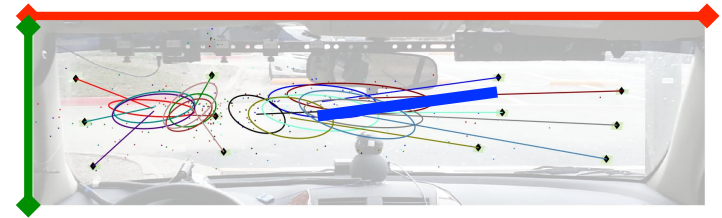
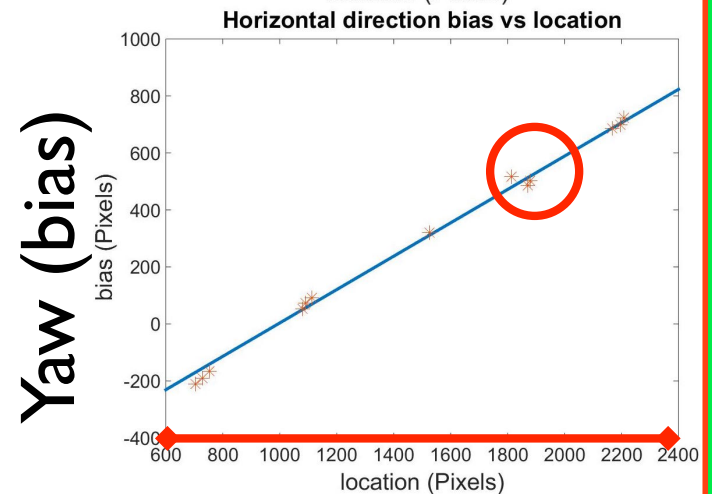
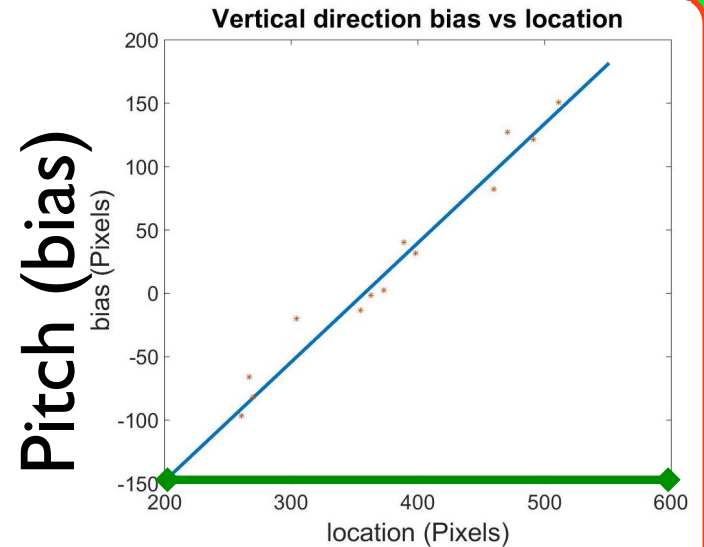
## Observations

- More variance (hence less predictability) when driving
- The length of the line represents the bias due to pupil movement between head pose and actual gaze
  - The bias increases as the direction moves away from the frontal pose
- Clear separation of gaze zones (front vs left)



# Head pose / gaze bias

- Bias increases for markers away from the center
  - Pupil movement is important
- Bias completely determines the vertical location
  - Slope in vertical direction is 1
  - No dependency on head pose





# Discussion

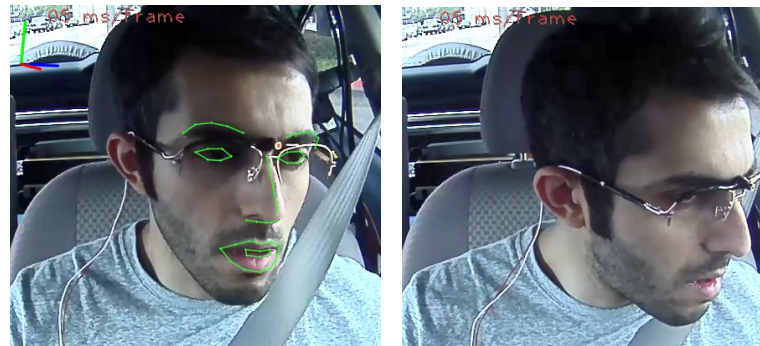
- High correlation in horizontal direction between head pose and gaze
- Low correlation in the vertical direction
- Most of the applications requires the knowledge of horizontal gaze (eg. blind spot, mirror checking etc.)
- Coarse estimation of gaze is possible using head pose which gives a general direction about visual attention





# Challenges

- Getting accurate head pose for highly skewed rotation



- Perspective mapping of point on the windshield to actual object in the road scene

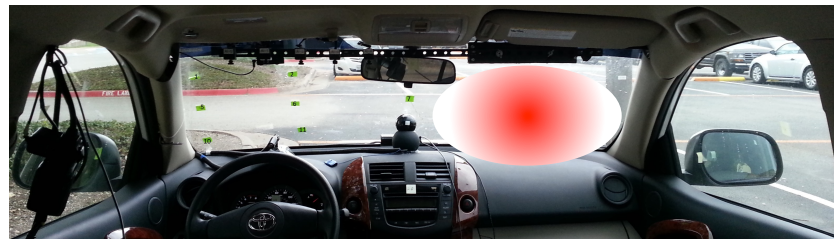




## Future work

- Use probabilistic modelling approach to get confidence level on various gaze location given a certain head pose

$$P(\text{gaze}(x, y) | \text{yaw}, \text{pitch}, \text{roll})$$



- Use road scene and contextual information for added confidence
- Extend the model to points not on the windshield (mirrors, windows, speedometer dial, radio etc.)



Thank you!  
Questions?

[MSP@UTDALLAS.EDU](mailto:MSP@UTDALLAS.EDU)