

EXPLORING FIELD OF VIEW IN HEAD MOUNTED DISPLAYS FOR MARITIME OBJECT DETECTION



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INTRODUCTION

Maritime *Search and Rescue (SAR)* activities involve missions to rescue people that are lost or in danger at sea. Today, still- and video-camera technologies e.g. mounted on drones, are increasingly being used to capture views that may not easily be accessible. To conduct user tests of different image presentations, we implement a *Head-Mounted Display (HMD) 360° Field-Of-View (FOV) simulator* to gain new insights into how to improve object detection by operators. Fig. 1. is a computer generated 360° panoramic maritime image. These scenes are often low observability - dark, low contrast, incorrect brightness, reflections off the water, weather – cloud, fog, rain, waves, sea spray. In addition, the objects may be far away and appear small. Objects of interest can include people, boats, ships and aircraft.

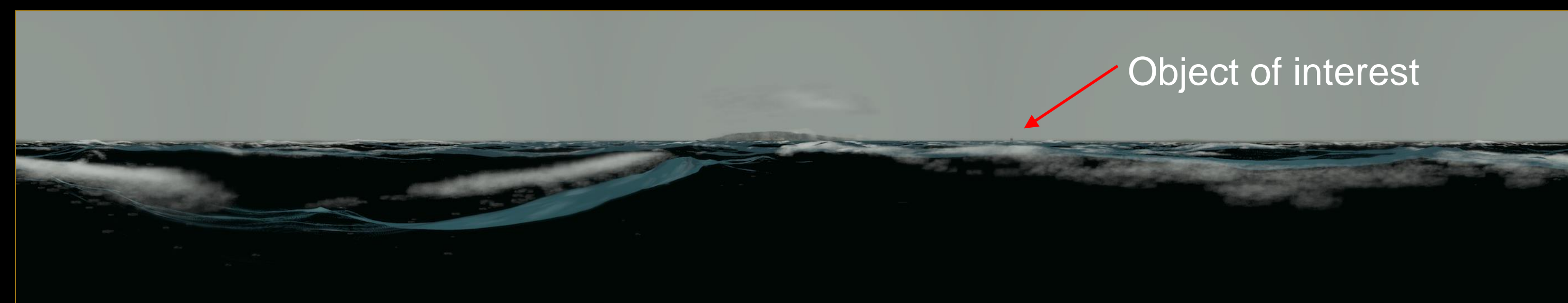


Figure 1. 360° panoramic maritime imagery. Operators perform an object detection task when viewing this imagery in a head-mounted display

AIM

Identify the most effective FOV and image representation (original, Canny, inverse) for object detection when viewing a maritime scene using a Head-Mounted Display.

METHODOLOGY

- 5 maritime images with objects of interest at 1km at random bearings (object represented by 4-6 pixels).
- 15s to detect object of interest.
- Head Mounted Displays:
 - i) HTC Vive¹ (resolution 2160x1200)
 - ii) Oculus Go² (resolution 2560 x 1440).
- 45 tests (5 images x 9 representations (3 FOVs x 3 filters)).
- 5 subjects.



Figure 2. User trial at the Curtin HIVE

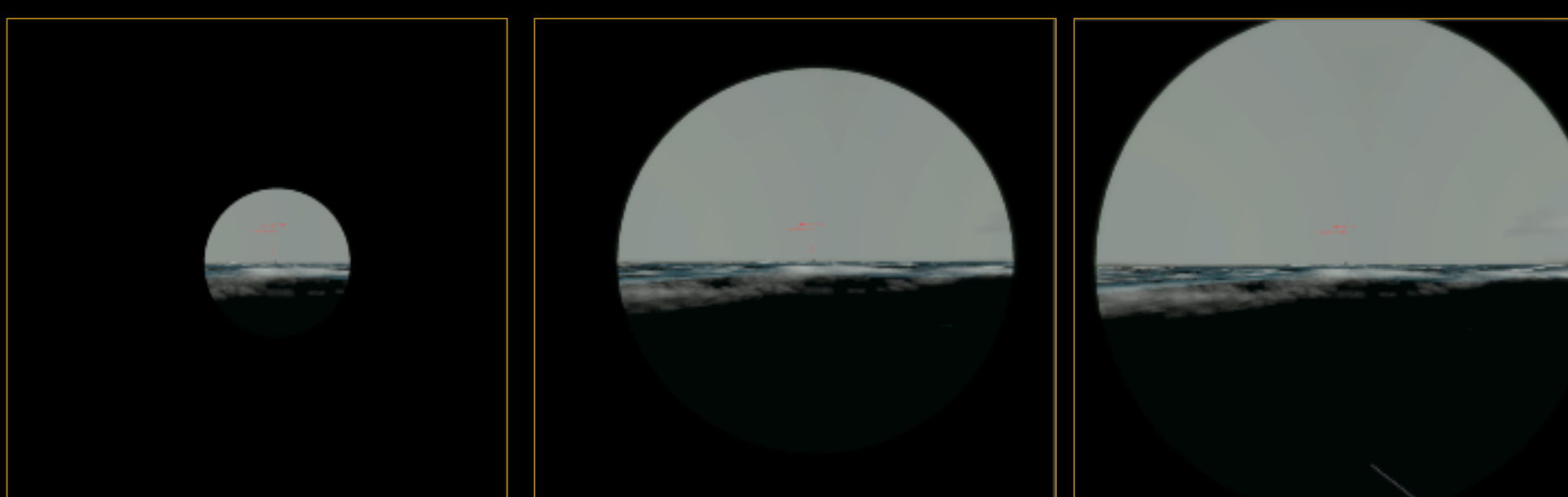
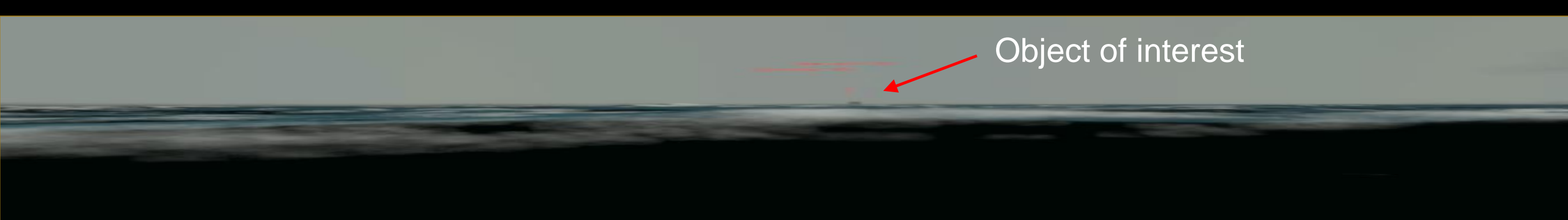
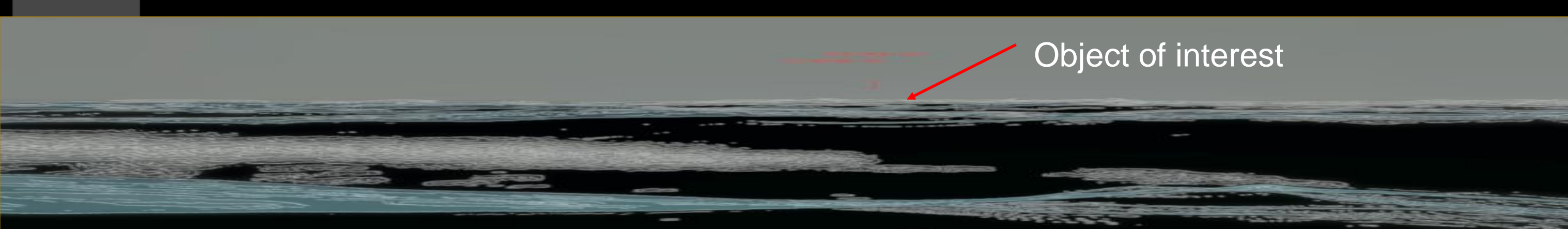


Figure 3. a) 30° FOV; b) 70° FOV; and c) 85° FOV



a) Original image



b) Canny filter



c) Inverse filter

Figure 4. Three image representations of the maritime scene: a) original; b) Canny; and c) inverse

INSTRUCTIONS FOR USING THE DEMO

- Experience the FOV simulator by putting on the Oculus Go HMD.
- One at a time, you will be shown 5 maritime images at different FOVs and image representation.
- Objects of interest are at a distance of 1km at random bearings.
- You have 15s to detect object of interest.
- Indicate to the demonstrator when you have detected the object of interest.
- There are 45 test images altogether (5 images x 9 representations (3 FOVs x 3 filters)).
- If you like, you can give us feedback via a NASA-TLX [1] survey.

METRICS

- Experiments were measured using the *NASA-TLX (Task Load Index)* [1]. NASA-TLX assesses mental demand, physical demand, temporal demand, performance, effort, frustration, time, accuracy and memory.
- High confidence self-perceived mental performance and physical demand were normalised.

RESULTS

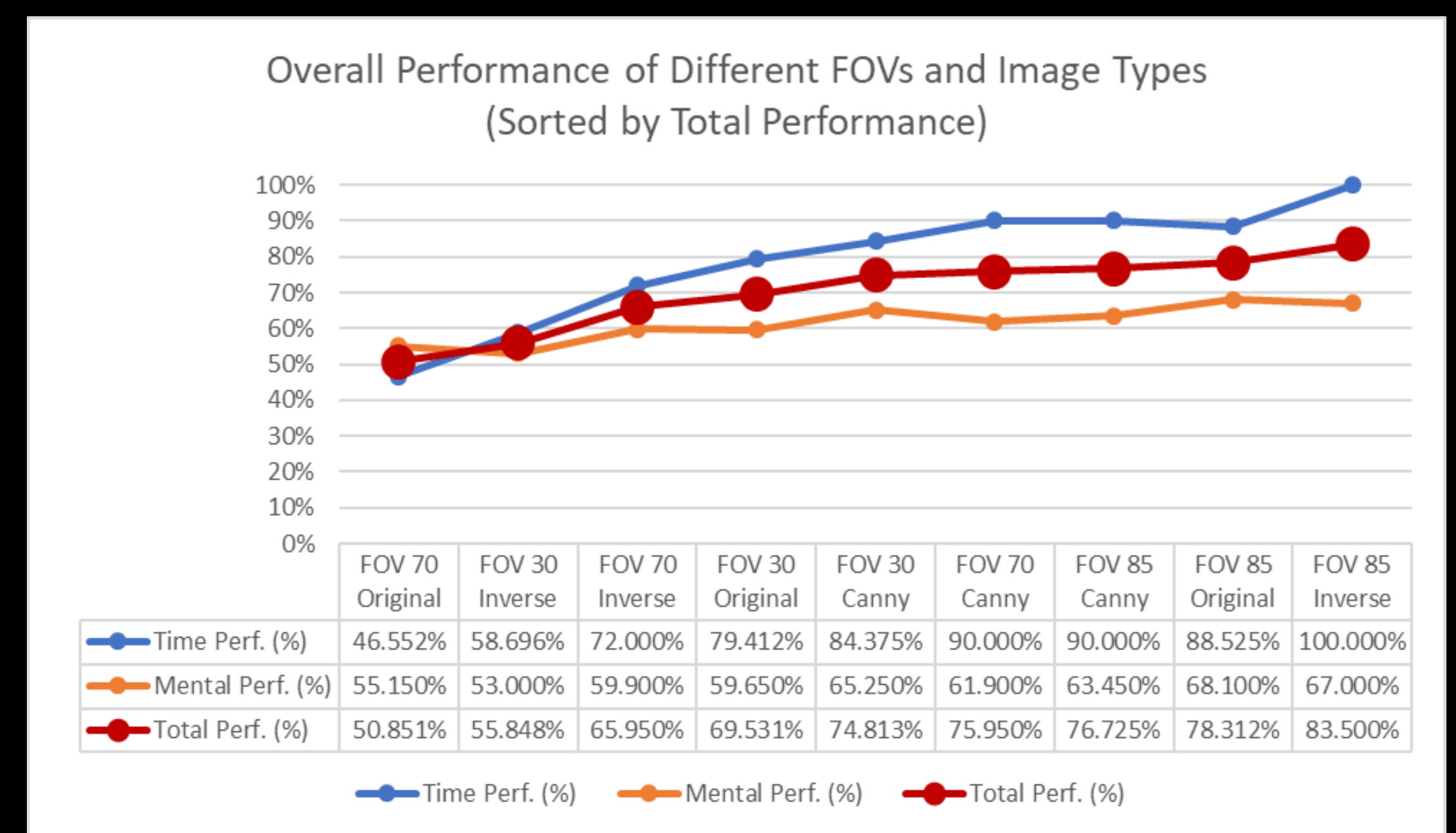


Figure 5. Overall performance of different FOVs and Image Types

- The wide FOV allowed users to perform the object identification task more accurately and quickly with both higher time and mental performance.
- Users observing the Canny image also performed well, where results for 30°, 70° and 85° FOVs are grouped together towards the right of Fig. 5.
- With the small 30° FOV, time and mental performance were relatively different for each image representation compared to other FOV values, meaning small FOVs are more sensitive to image representation.

DISCUSSION

- Object detection relies on first the peripheral vision to scan the scene and then central vision to perform the object identification.
- Wide FOVs give the user more immediate information to analyse the scene quickly before achieving object identification.
- The Canny image was most successful, isolating areas in the image to improve object detection.
- The inverse image was also successful, offering segmentation of the image into areas; it was lower contrast and had less clutter and noise, for better object detection.

ACKNOWLEDGMENT

We thank Richard Palmer, Ryan Day and Jesse Helliwell (Curtin HIVE) for programming.

REFERENCES

[1] Hart, S. G. and Staveland, L. E., Development of NASA-TLX (Task Load Index): Results of empirical and theoretical research, *Advances in Psychology*, Vol. 52, 139-183, 1998.

Tiled

Cylinder

Wedge

Dome