

### **Introduction to Intelligent User Interfaces**

**Context Awareness Interaction in Smart Environments** 

"A SONY 4K TV' NORMAN HARVEY CONNECTED HOME SHOWHOUSE [IDEAL HOME SHOW 17th - 19th April 2015] - 103590" by infomatique is licensed with CC BY-SA 2.0



### **Context Awareness Interaction**



Karan Ahuja, Andy Kong, Mayank Goel, and Chris Harrison. 2020. Direction-of-Voice (DoV) Estimation for Intuitive Speech Interaction with Smart Devices Ecosystems. In Proceedings of the 33rd Annual ACM Symposium on User Interface Software and Technology (UIST '20). ACM, New York, NY, USA, 1121–1131. DOI: <u>https://doi.org/10.1145/3379337.3415588</u>

### **Breakout Sessions**

- List as many addition information which would support am intelligent user interface?
- Scenario 1: Touch Screen Interaction
- Scenario 2: Navigation
- Scenario 3: Movie
- Scenario 4: Voice activated lights in a living room



# **Extracting Contextual Information**

- Users Location
  - GPS
  - Direction of Voice
- Users Activity
- Users Emotion
- Users Pose
- Objects Surrounding the User
- Status of Objects

### **Smart Environments - Direction-of-Voice**

- Feature Extraction e.g. volume, speech frequency ratio
- Machine Learning e.g. ensemble-based decision trees



Karan Ahuja, Andy Kong, Mayank Goel, and Chris Harrison. 2020. Direction-of-Voice (DoV) Estimation for Intuitive Speech Interaction with Smart Devices Ecosystems. In Proceedings of the 33rd Annual ACM Symposium on User Interface Software and Technology (UIST '20). ACM, New York, NY, USA, 1121–1131. DOI: <u>https://doi.org/10.1145/3379337.3415588</u>

### Users' Context in Smart Environments

## **Smart Environments - Direction-of-Voice**



Karan Ahuja, Andy Kong, Mayank Goel, and Chris Harrison. 2020. Direction-of-Voice (DoV) Estimation for Intuitive Speech Interaction with Smart Devices Ecosystems. In Proceedings of the 33rd Annual ACM Symposium on User Interface Software and Technology (UIST '20). ACM, New York, NY, USA, 1121–1131. DOI: <u>https://doi.org/10.1145/3379337.3415588</u>

#### Users' Context in Smart Environments

# **Context-Aware Keyboards**



- Palm detection input rejection
- Finger identification, Finger Orientation model improvement
- Grip detection model selection

# **Palm Detection**

- Convolutional Neural Network
- Classification
- Representation Learning





Huy Viet Le, Thomas Kosch, Patrick Bader, Sven Mayer, and Niels Henze. 2018. PalmTouch: Using the Palm as an Additional Input Modality on Commodity Smartphones. In Proceedings of the 2018 CHI Conference on Human Factors in Computing Systems (CHI '18). ACM, New York, NY, USA, Paper 360, 1–13. DOI: <u>https://doi.org/10.1145/3173574.3173934</u>

# **Finger Orientation**



Sven Mayer, Huy Viet Le, and Niels Henze. 2017. Estimating the Finger Orientation on Capacitive Touchscreens Using Convolutional Neural Networks. In Proceedings of the 2017 ACM International Conference on Interactive Surfaces and Spaces (ISS '17). ACM, New York, NY, USA, 220–229. DOI: <u>https://doi.org/10.1145/3132272.3134130</u>

### Users' Context in Smart Environments

### **Activity Recognition – Accelerometers**



Mi Zhang and Alexander A. Sawchuk. 2012. USC-HAD: a daily activity dataset for ubiquitous activity recognition using wearable sensors. In Proceedings of the 2012 ACM Conference on Ubiquitous Computing (UbiComp '12). Association for Computing Machinery, New York, NY, USA, 1036–1043. DOI: <u>https://doi.org/10.1145/2370216.2370438</u>

#### Users' Context in Smart Environments

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# **Neuronal Network With Timeseries Data**

**Recurrent Neural Network** 



In depth LSTM tutorial: <u>http://colah.github.io/posts/2015-08-Understanding-LSTMs/</u> Code Examples: <u>https://github.com/cwi-dis/mobile-har-tutorial</u>

#### Users' Context in Smart Environments

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### **Human Pose Detection**

- Keypoint Estimation
- Multi-stage CNN



Zhe Cao, Gines Hidalgo, Tomas Simon, Shih-En Wei, and Yaser Sheikh. OpenPose: realtime multi-person 2D pose estimation using Part Affinity Fields. IEEE transactions on pattern analysis and machine intelligence 43, no. 1 (2019): 172-186. DOI: <u>https://doi.org/10.1109/TPAMI.2019.2929257</u>

#### Users' Context in Smart Environments

### **Camera based Gaze Estimation**











Projection on screen



Facial landmarks

Gaze prediction

Gaze direction



Xucong Zhang, Yusuke Sugano, and Andreas Bulling. 2019. Evaluation of Appearance-Based Methods and Implications for Gaze-Based Applications. In Proceedings of the 2019 CHI Conference on Human Factors in Computing Systems (CHI '19). ACM, New York, NY, USA, Paper 416, 1–13. DOI: <u>https://doi.org/10.1145/3290605.3300646</u> URL: <u>http://www.opengaze.org/</u>

#### Users' Context in Smart Environments

### **Camera based Gaze Estimation**

- Convolutional Neural Network
- Regression x/y cornindates



Xucong Zhang, Yusuke Sugano, Mario Fritz, and Andreas Bulling. 2017. It's written all over your face: Full-face appearancebased gaze estimation. In Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition Workshops, pp. 51-60. 2017. DOI: <u>https://doi.org/10.1109/CVPRW.2017.284</u>

#### Users' Context in Smart Environments

## **Enhanced Voice Assistants**



Front Camera

Rear Camera



Sven Mayer, Gierad Laput, and Chris Harrison. 2020. Enhancing Mobile Voice Assistants with WorldGaze. In Proceedings of the 2020 CHI Conference on Human Factors in Computing Systems (CHI '20). Association for Computing Machinery, New York, NY, USA, 1–10. DOI: <u>https://doi.org/10.1145/3313831.3376479</u>

#### Users' Context in Smart Environments

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### **Object Detection**



Sven Mayer, Gierad Laput, and Chris Harrison. 2020. Enhancing Mobile Voice Assistants with WorldGaze. In Proceedings of the 2020 CHI Conference on Human Factors in Computing Systems (CHI '20). Association for Computing Machinery, New York, NY, USA, 1–10. DOI: <u>https://doi.org/10.1145/3313831.3376479</u>

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# **Object Detection**

**R-CNN: Regions with CNN feature** 



Ross Girshick, Jeff Donahue, Trevor Darrell, and Jitendra Malik. Rich feature hierarchies for accurate object detection and semantic segmentation (2014) In Proceedings of the IEEE conference on computer vision and pattern recognition, pp. 580-587. DOI: <u>https://doi.org/10.1109/CVPR.2014.81</u> Source Code: <u>https://github.com/rbgirshick/rcnn</u>

# **Object Detection**

YOLO – You Only Look Once



### Users' Context in Smart Environments

## Machine Learning for HCI



### Adaptation for Machine Learning



Huy Viet Le, Sven Mayer, and Niels Henze. 2020. Deep learning for human-computer interaction. interactions 28, 1 (January -February 2021), 78-82. DOI: https://doi.org/10.1145/3436958



gens pranning enables despisaring enables despisaring solutions by device approaches. Determining a solution's qualities in easy for most disciplines but much ander for ROL, as it needs to conduct a users study. The user-centered design for users and the adopted by incorporating data disciplicition and iterative model devicement.	has truey long realistics in human- compary in succession (HCL). Instable human-centred processes are widdly used in a advantia and industry to the success and the succession of the human structure of the succession of the succession human structure of the succession human structure of the succession human structure of the succession human structure of the succession human scentered design	work in many disciplicus, such as computer vision, narrul language processing, and brain machine interfaces. These disciplications want fitness approaches for building novel systems. In expensive iterative processes, machine learning models are trained and fine- stund, shalts is possible only constrant to FICL, such models typically do not follow human-centrered approaches, as the developed solution can be evaluated using ingeline metrics, the quality of a solution, which turns
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#### Users' Context in Smart Environments

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# **Neuronal Network Concepts**

Introduced concepts

- Classification vs. Regression
- Feature Extraction vs. Representation Learning
- Model Structures
  - Neuronal Network
  - Convolutional Neural Network
  - RNN
  - Long short-term memory (LSTM)