## BatVision with GCC-PHAT Features for Better Sound to Vision Predictions

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#### **Existing Sensors have some Drawbacks**

- Vision is valuable sensor but sometimes fails
- Ultrasound, Radar or LIDAR sensors are often costly, complex and provide limited information





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#### **Approach Inspired by Nature**



Image credit: https://askabiologist.asu.edu/echolocation











Chirp:

• 3 Milliseconds



#### **Neural Network Prediction of Visual Layout from Sound**



#### **Dataset: Binaural echo to depth**



#### **Data Collection in Office Building**



#### Echos cut off at fixed distance



#### **Structure Beyond can not be Directly Observed**







## **Architecture Overview** Binaural Ground truth Base generator on residual learning Adversarial loss Discriminator А ( ¬ L1 Regression loss Audio Encoder Generator

#### **Architecture Overview**



[1] Isola et al., Image-to-Image Translation with Conditional Adversarial Networks,

https://arxiv.org/pdf/1611.07004.pdf Christensen, Hornauer & Yu, BatVision with GCC-PHAT Features for Better Sound to Vision Predictions



Grayscale image from camera



Grayscale image from camera

Depth map from stereo camera



Grayscale image from camera

Depth map from stereo camera

Predicted depth map from BatVision [1]



Grayscale image from camera

#### Depth map from stereo camera

Predicted depth map from BatVision [1] Predicted depth map from our improved BatVision model







#### Scene













Scene Stereo GT



















Scene Stereo GT BatVision [1]





















Scene Stereo GT BatVision [1] Ours

# Results: Grayscale layout No depth used for training!

Plausible layout of free space / obstacles



#### Scene



# Results: Grayscale layout No depth used for training!

Plausible layout of free space / obstacles



Scene BatVision [1]

Scene BatVision [1]

# Results: Grayscale layoutNo depth usedPlausible layout of free space / obstaclesfor training!

Scene BatVision [1] Ours

Scene BatVision [1]

Ours

### Conclusion

- ➔ Improving sound-to-vision
- → Increasing perceptual quality and quantitative measures
- → More stable training process
- → Less noisy depth and layout predictions

