## SOUND LOCALIZATION AND RECOGNITION





#### Agenda

- Sound localization in humans
- Binaural methods
- Array methods
- Humanoid robot approach

## Sound Localization in Humans



#### Sound perception in humans

- The outer ears collect sound and amplify them
- The inner ears use the changes in air pressure to convert them into neural signals
- Fluid elements in the ear used to detect frequency

#### Anatomy of the Ear



#### Medial Superior Olive

- Receives input from both ears AVCN's(anteroventral cochlear nucleus)
- Calculates low frequency sounds
- Interaural time difference

#### Lateral Superior Olive

- Receives input from both ears AVCN's(anteroventral cochlear nucleus)
- Calculates high frequency sounds
- Interaural level difference



#### Sound Localization

- Monaural
- Vertical processing
- Binaural
- Processed in the MNTB(Medial nucleus of the trapezoid body)
- Horizontal processing

#### **Interaural Time Difference**

- Differences in travel time from a sound source to closer and farther ear
- Mathematically works well for frequencies
  >=2kHz
- In humans highest sensitivity between 0.7 and 1 kHz, insensitive after ~1.4 kHz

#### Interaural Level Difference

- For smaller wavelengths (higher frequencies)
- Physical presence of the head makes a shadow
- Louder in one ear than the other



## Sound Localization in Robotics



#### **Potential Constraints**

- Geometric
- Embeddability
- Real time
- Broadband
- Environmental

## Approach

- Frequency decompositions modelled by FFT, bandpass filters or gammatone filters
- Binaural processing
- Array processing

#### **Binaural Processing**

- Simpler computations
- Simpler geometry
- Built in irregularity



#### Head Related Transfer Function

The HRTF captures the relationship between the signal originating from a sound source and captured at a certain arbitrary reference position in space

- Auditory Epipolar Geometry
- Revised Auditory Epipolar Geometry
- Scattering Theory

 $\begin{cases} L(f) = H_L(r_s, \theta_s, \varphi_s, f)S(f), \\ R(f) = H_R(r_s, \theta_s, \varphi_s, f)S(f), \end{cases}$ 



#### **Adaptive Filters**

- Robots own noise
- Environmental reflection
- Unexpected environmental elements



#### Neural Networks

- Learning head effect
- Learn different environments



#### **Gaussian Models**

- Auditory events with ITD values in histograms
- Approximated with gaussian models
- Results in model with peaks from sound sources
- Can compute multiple sound sources



https://www.semanticscholar.org/paper/Gaussian-process-models-for-HRTF-based-3D-sound-Luo-Zotkin/24f37465aae38112c3ad5969d786e47a07b4824d

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#### **Spectral Cues**

- Elevation dependent on positions of notches in the spectrum of sound signals
- Acoustic reflections from head and outer ear
- Scattering patterns from shape of pinnae

#### **Distance Localization**

- Triangular theory
- Theoretically only for static sources
- Also not too accurate in humans





https://www.researchgate.net/figure/Triangular-configuration-of-sensor-nodes-sound-source-for-the-localization-where-Pi\_fig5\_236006629



#### Array processing

- Using array of microphones
- Computation heavy
- Mostly not broadband capable



#### MUSIC

- MUltiple SIgnal Classification
- Assumes sound source is independent, zero mean stationary and of a single frequency
- Can compute multiple frequencies discretely



#### 

#### TDOA

- Time Delay(s) of Arrival
- Calculate delay between two microphones of array
- Similar notion to ITD, but different computation

#### Beamforming

- Most used in robotics
- Low computation cost
- Performance dependent on number microphones
- Computes energy map



#### Humanoid Robots

- We look at one proposed algorithm for sound localization in humanoid robots
- Proposed to handle reverberance
- Localize sound in 3d space
- Potential real time calulations

#### STATE-OF-THE-ART

- Uses spatial likelihood functions of several microphone arrays
- These are then integrated with a weighted average of each function
- Uses multiple possible results to find the best one
- Can be done probabilistically or with neural networks

#### Monaural Algorithm

- Each ear has two microphones, one inside the ear and one outside
- A spatially modified acoustic signal received inside the ear
- Echoes and noise are modeled for
- Position identified by filter response on the noise from look up table



#### **Binaural Algorithm**

- Using the inner ear input of each ear
- If we divide the two inputs, to get the ratios we also cancel the source
- We match this to lookup table of HRTF pairs



## **Unified Sound Localization System**

- Combine results of each ear monaural algorithm and the binaural algorithm
- If they are between 5 degres of each other they are averaged
- Else the angular error is calculated
- Bayesian network used to detect the direction
- The robot then decides the status of sound event



#### Conclusion

- The system cannot calculate distance
- Cannot localize multiple sound sources
- Does not have auditory environment
- Potentially can use head movements or robot repositioning

#### References

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- Chapter-6---Sound-localization\_2015\_Handbook-of-Clinical-Neurology
- A survey on sound source localization in robotics: From binaural to array processing methods, S.Argentieria, P.Danès, P.Souères
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# Questions

