

# ADVANCED EAR EXAMINATION USING DEEP LEARNING-ASSISTED MOBILE OTOSCOPE

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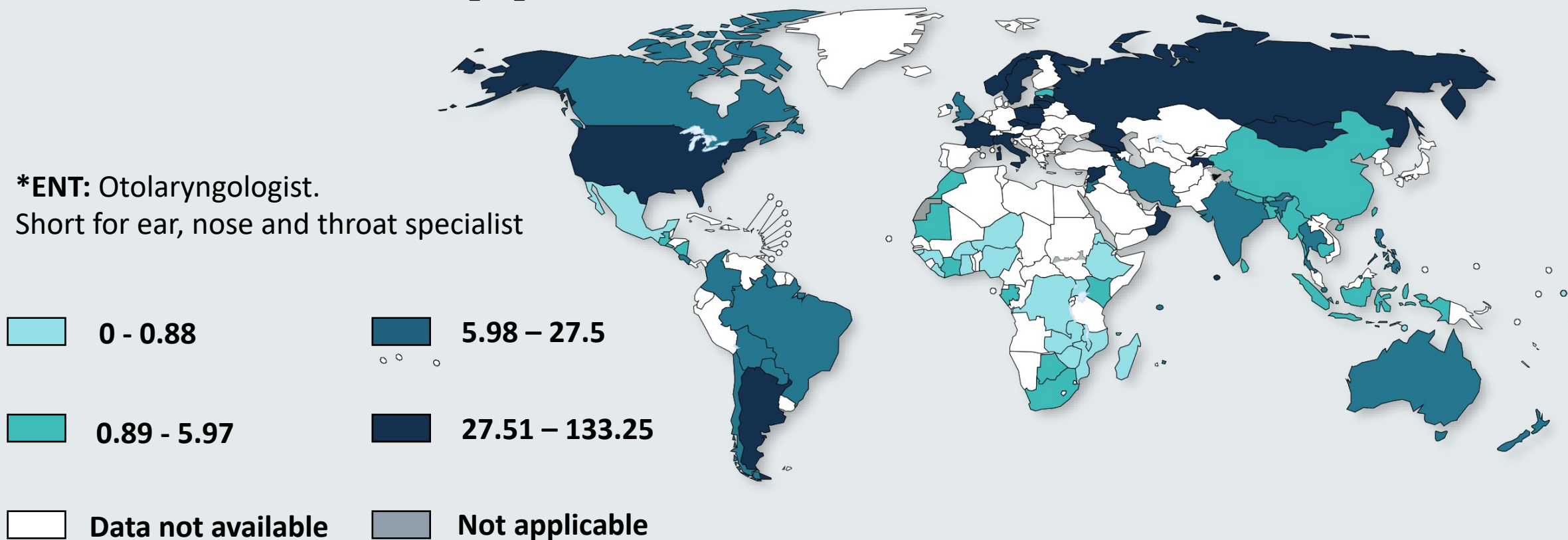
**2. Research approach**

**3. Validation**

**4. Conclusion**

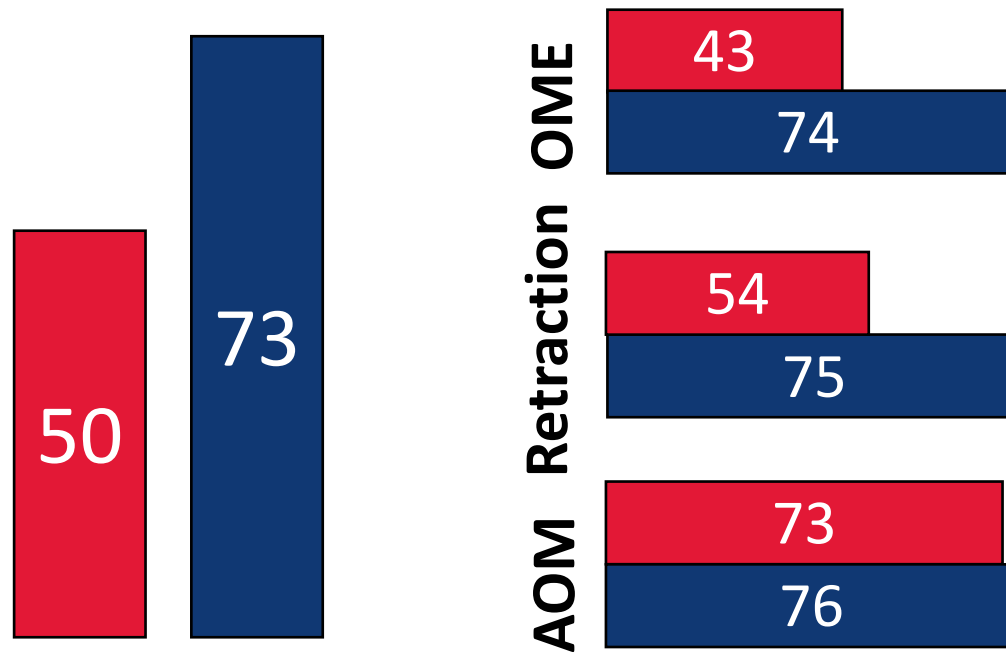
## Number of ENT specialists per million population, worldwide in 2013 [1]

\*ENT: Otolaryngologist.  
Short for ear, nose and throat specialist



[1] World Health Organization. "Multi-country assessment of national capacity to provide hearing care." (2013).

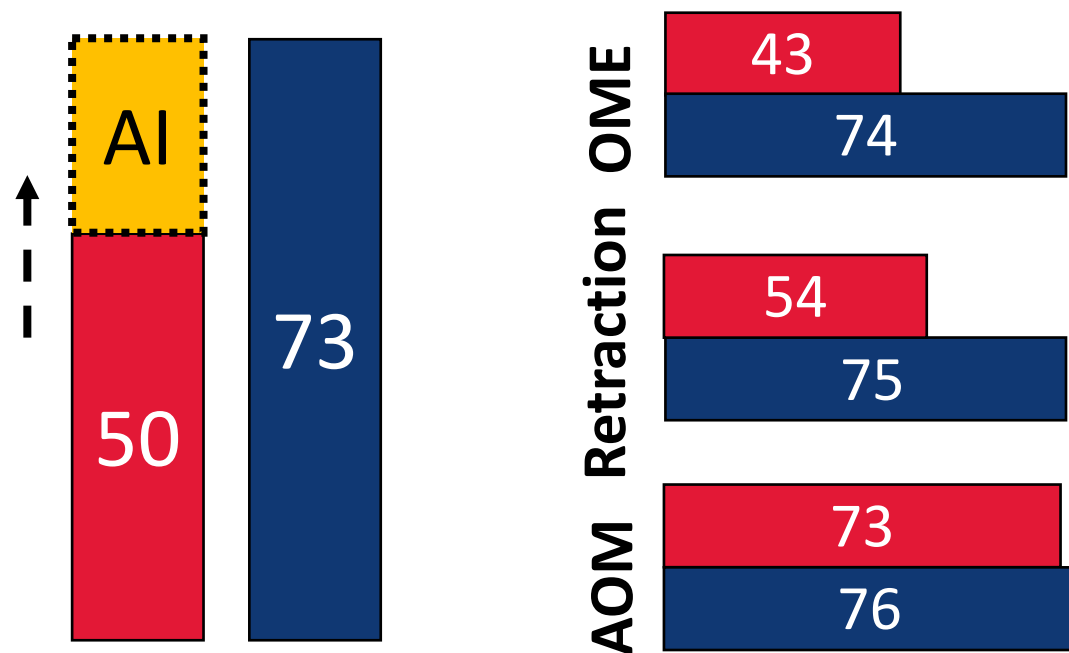
## Comparison of diagnostic accuracy (%) between **pediatrician** and **ENT specialist** [2]



\***OME**: Otitis media with effusion, **AOM**: Acute otitis media

[2] Pichichero, Michael E., and Michael D. Poole. "Assessing diagnostic accuracy and tympanocentesis skills in the management of otitis media." Archives of pediatrics & adolescent medicine 155.10 (2001): 1137-1142.

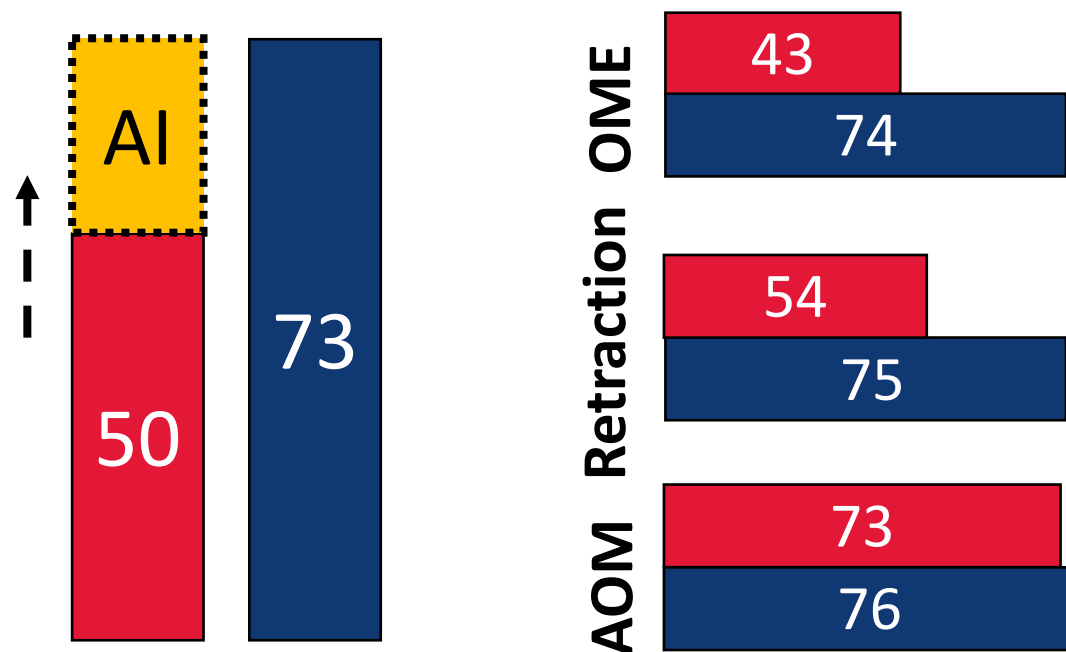
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We suggest development of **non-specialist and affordable** ear examination tool

## **2. Research approach**

# Deep learning for otolaryngology

Study	Application	Accuracy
Basaran (2020) [3]	Diagnosis of middle ear inflammation	<b>90.48%</b>
Cha (2019) [4]	Detection of ear and mastoid disease	<b>93.67%</b>
Livingstone (2019) [5]	Otologic disease screening	<b>84.4%</b>

✓ **Convolutional neural networks** are reported **strong** performance

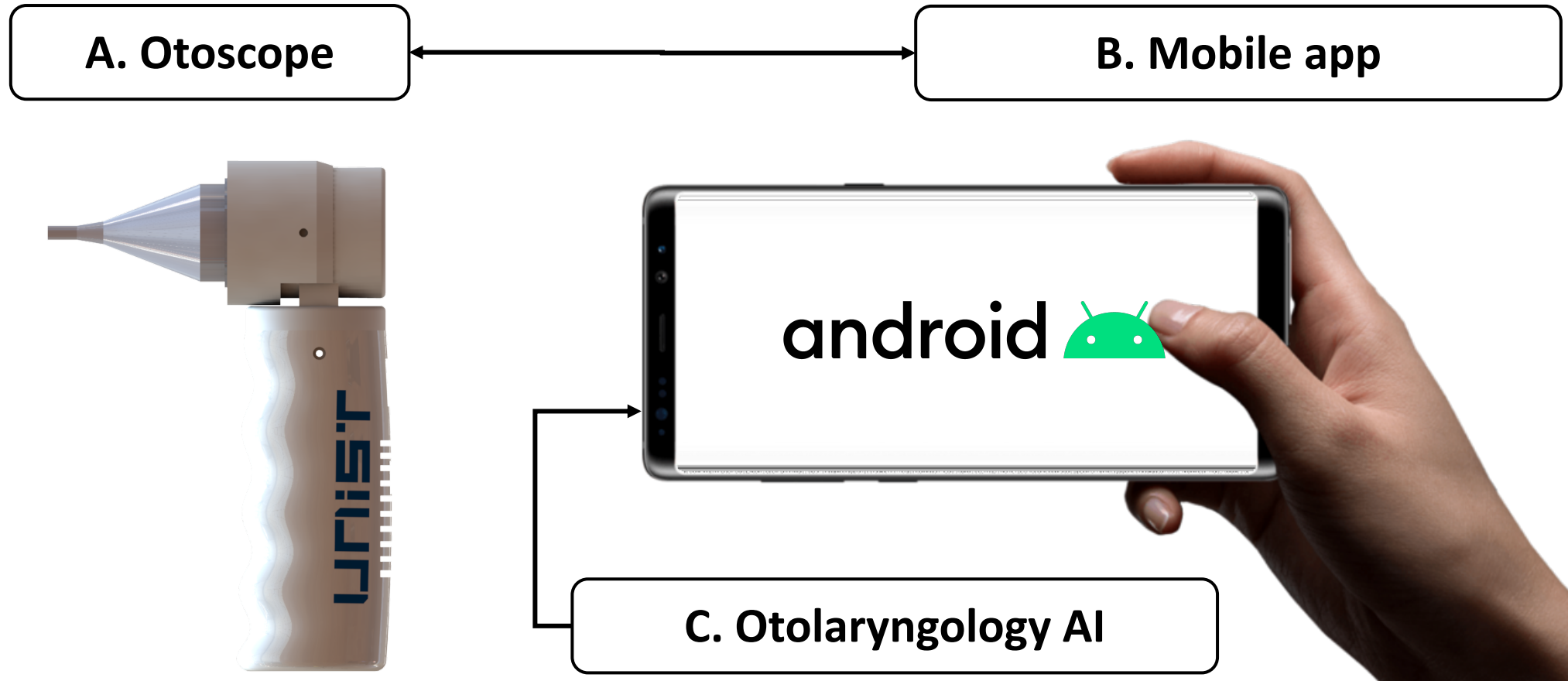
[3] Cha, Dongchul, et al. "Automated diagnosis of ear disease using ensemble deep learning with a big otoendoscopy image database." *EBioMedicine* 45 (2019): 606-614.

[4] Livingstone, Devon, et al. "Building an Otoscopic screening prototype tool using deep learning." *Journal of Otolaryngology-Head & Neck Surgery* 48.1 (2019): 1-5.

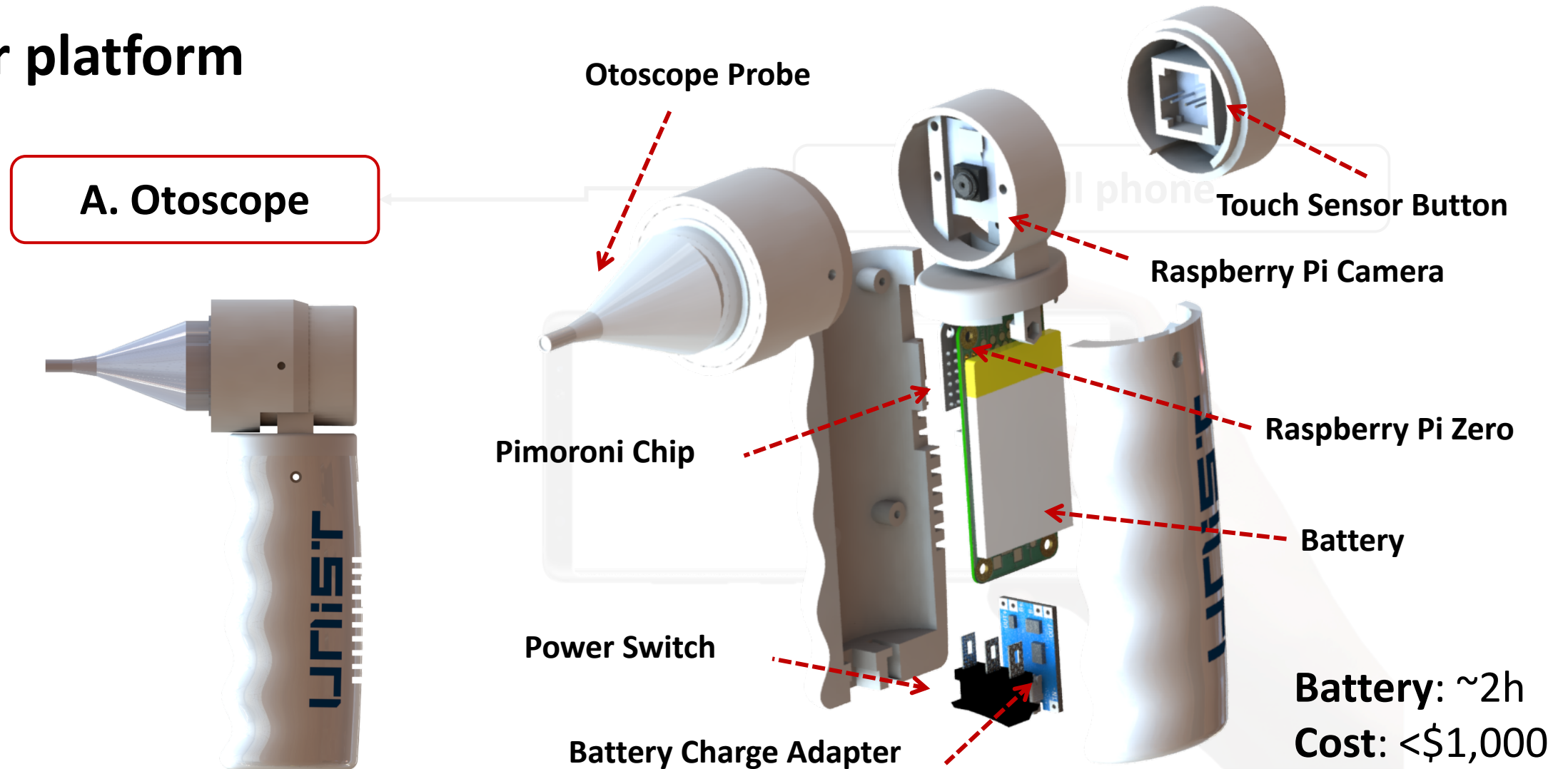
[5] Başaran, Erdal, Zafer Cömert, and Yüksel Çelik. "Convolutional neural network approach for automatic tympanic membrane detection and classification." *Biomedical Signal Processing and Control* 56 (2020): 101734.



## Our ear examination platform



# Our platform



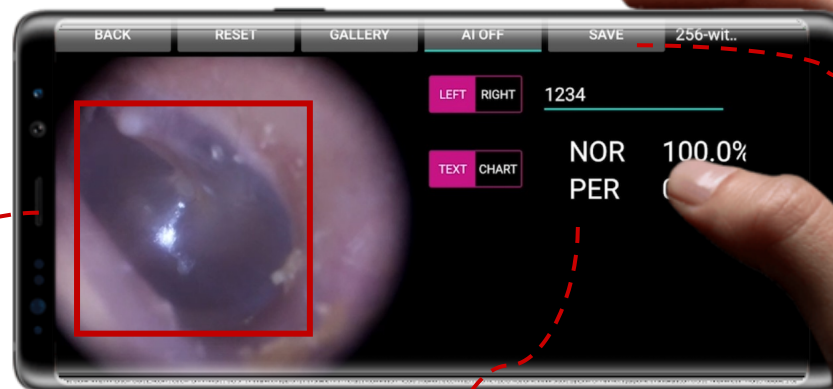
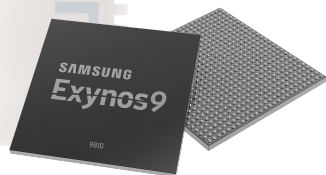
# Our platform

A. Otoscope

B. Mobile app

Custom developed  
Android application

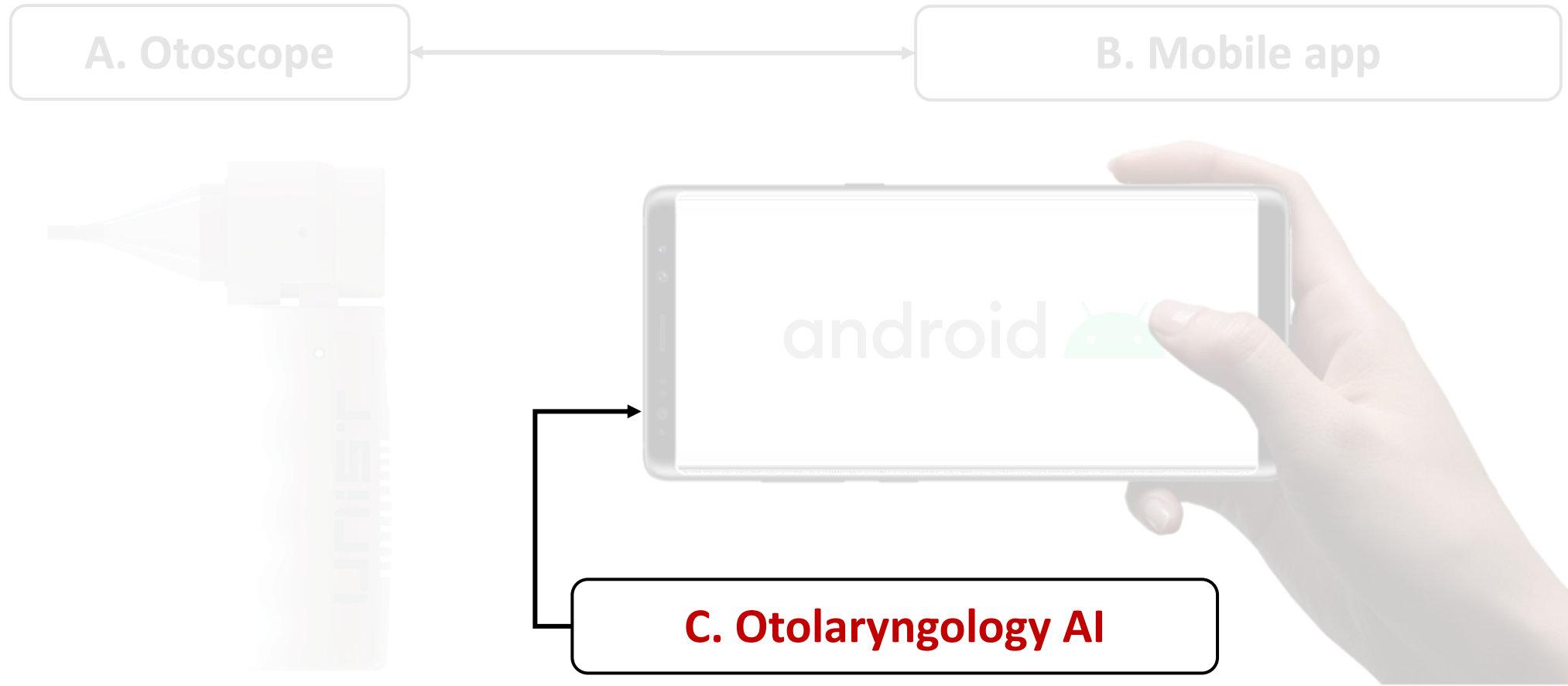
✓ A. Wireless Stream:  
<100ms low-latency



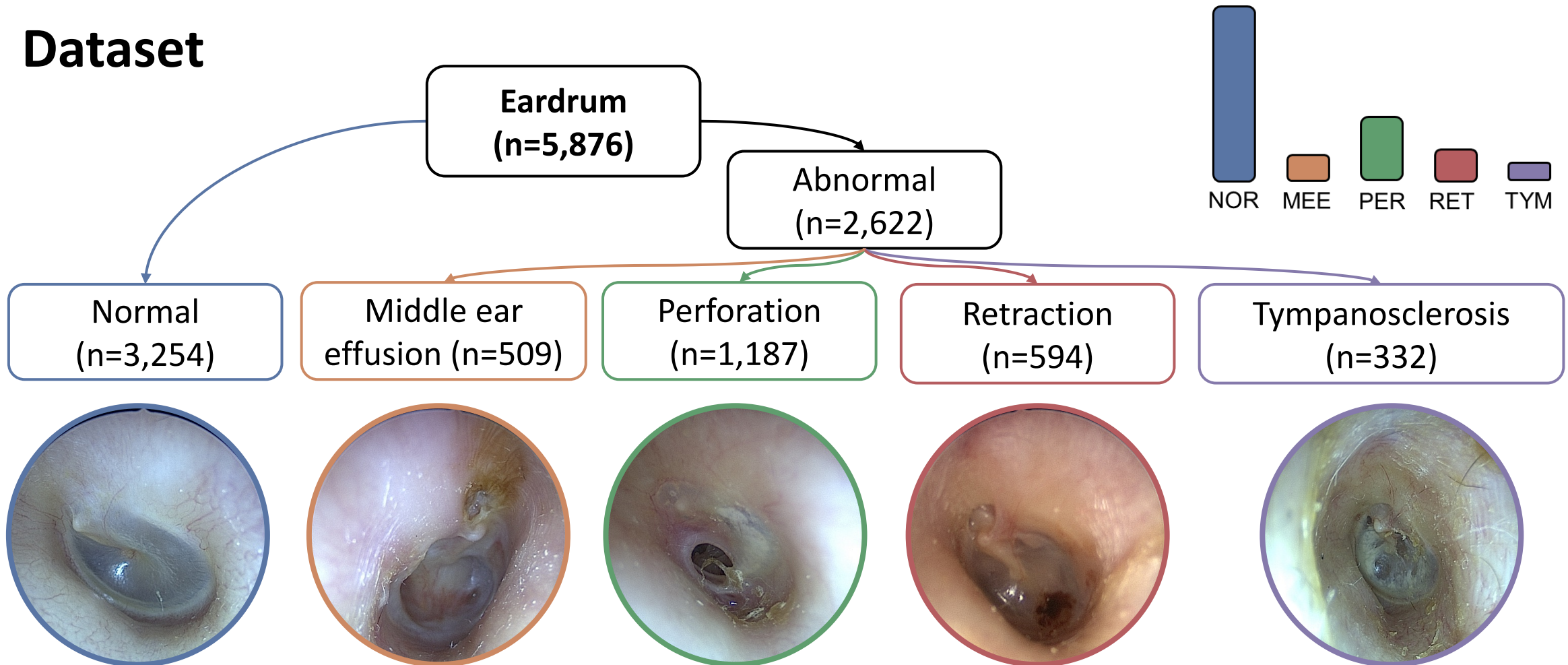
Tested on Galaxy S9+, S10+, Note

✓ C. Real-time AI inference on  
mobile GPU <10ms/image

# Our platform

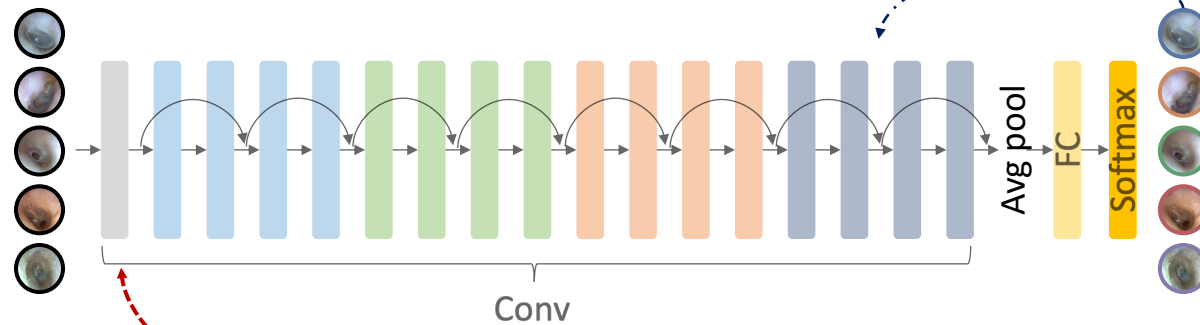


# Dataset



Received from Department of Otorhinolaryngology, Ajou University Hospital, IRB (AJIRB-MED-OBS-21-409)

## Dataset

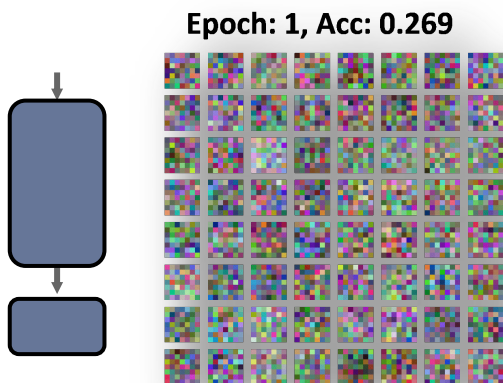


Revealing classification using Grad-CAM [7]

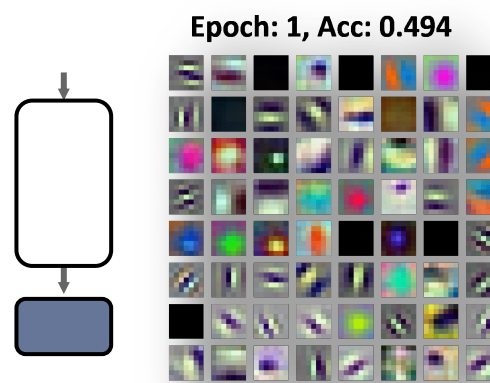
$$L_{Grad-CAM}^c = ReLU \left( \frac{1}{Z} \sum_k \sum_i \sum_j A^k \frac{\partial y^c}{\partial A_{i,j}^k} \right)$$

\*Linear combination; Global average pooling; Backprop gradients

Extracting features using ResNet-18 [6]

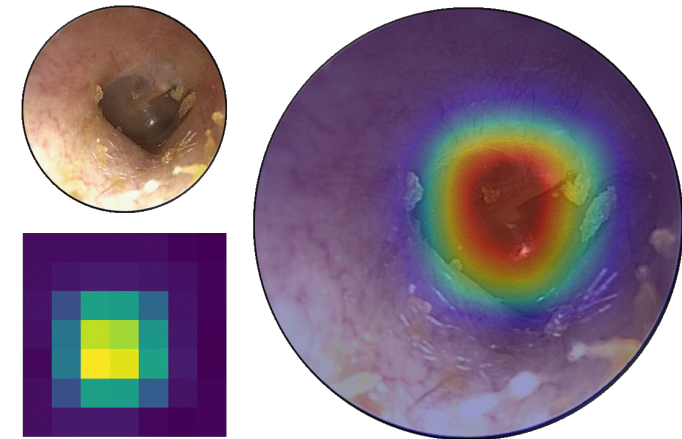


Training entire model



Transfer learning

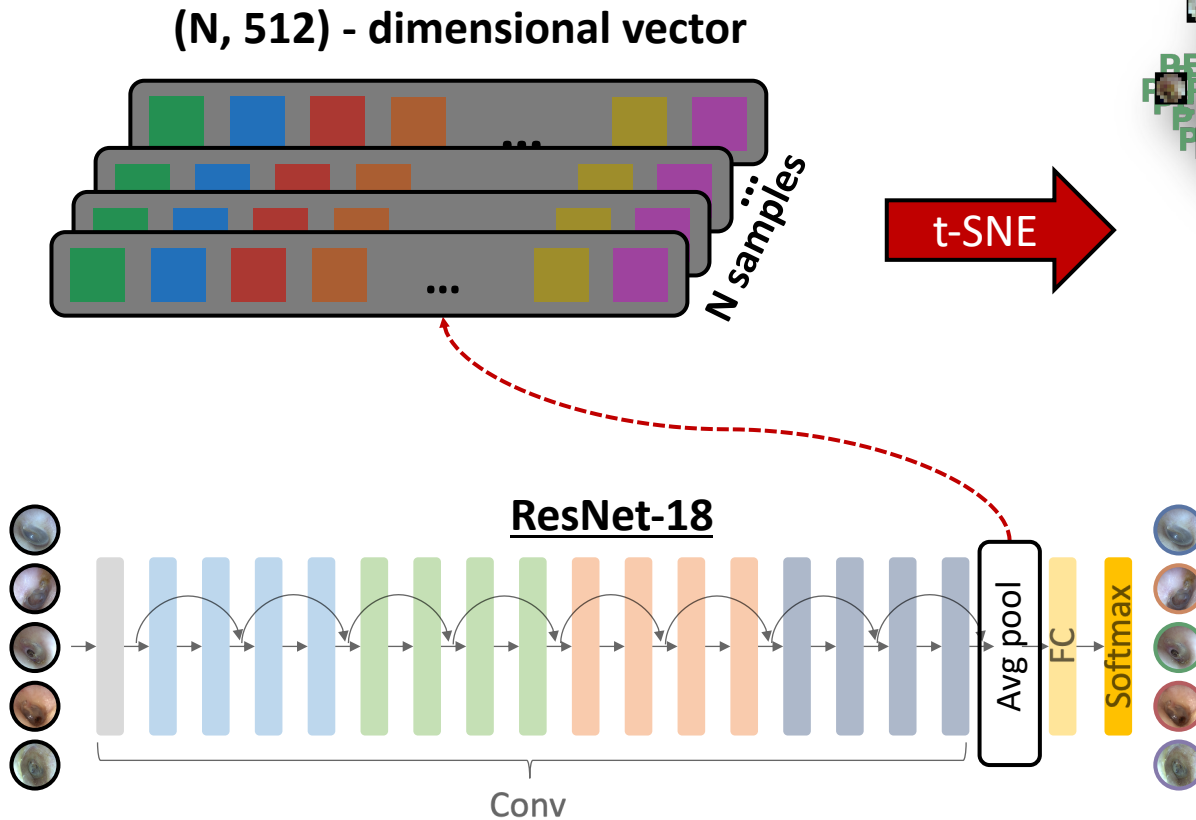
**NORMAL**



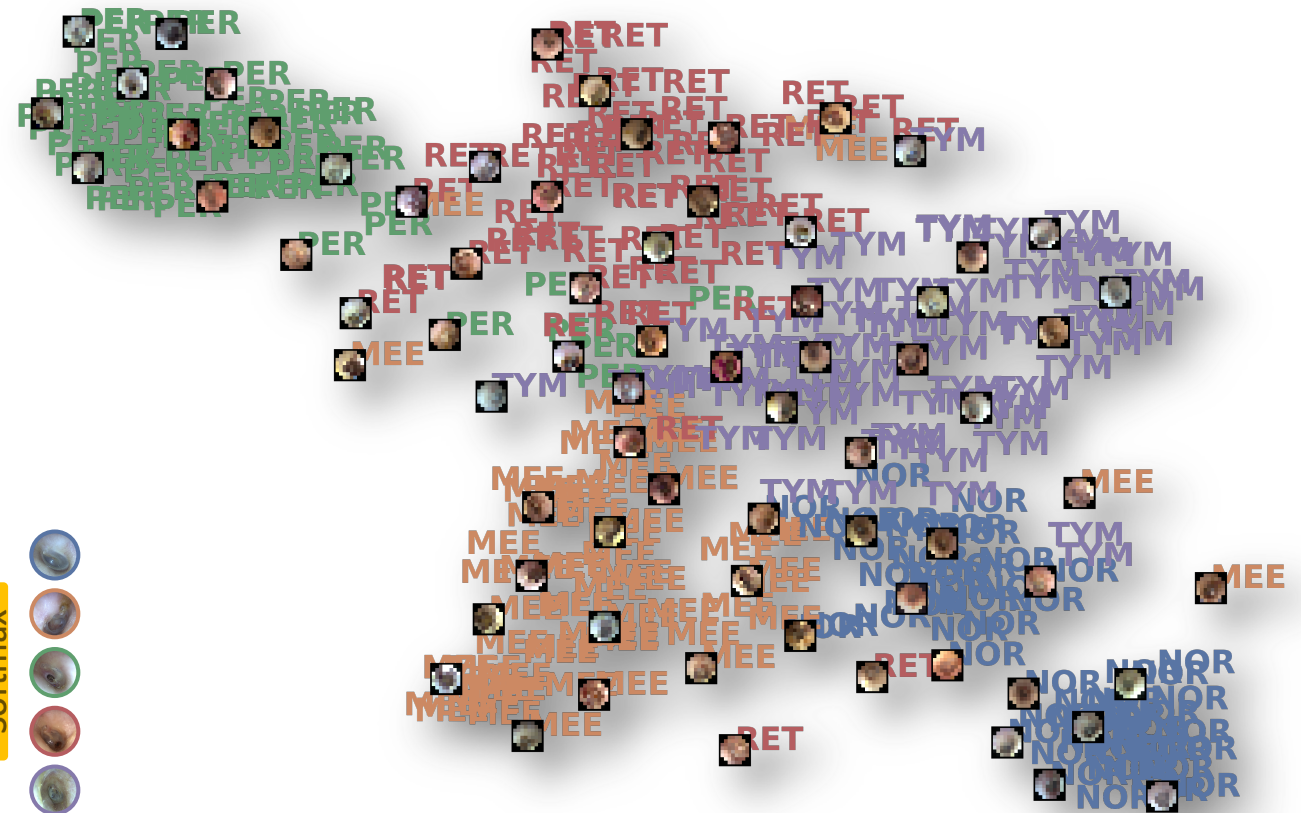
[6] He, Kaiming, et al. "Deep residual learning for image recognition." *Proceedings of the IEEE conference on computer vision and pattern recognition*. 2016.

[7] Selvaraju, Ramprasaath R., et al. "Grad-cam: Visual explanations from deep networks via gradient-based localization." *Proceedings of the IEEE international conference on computer vision*. 2017.

# Dataset

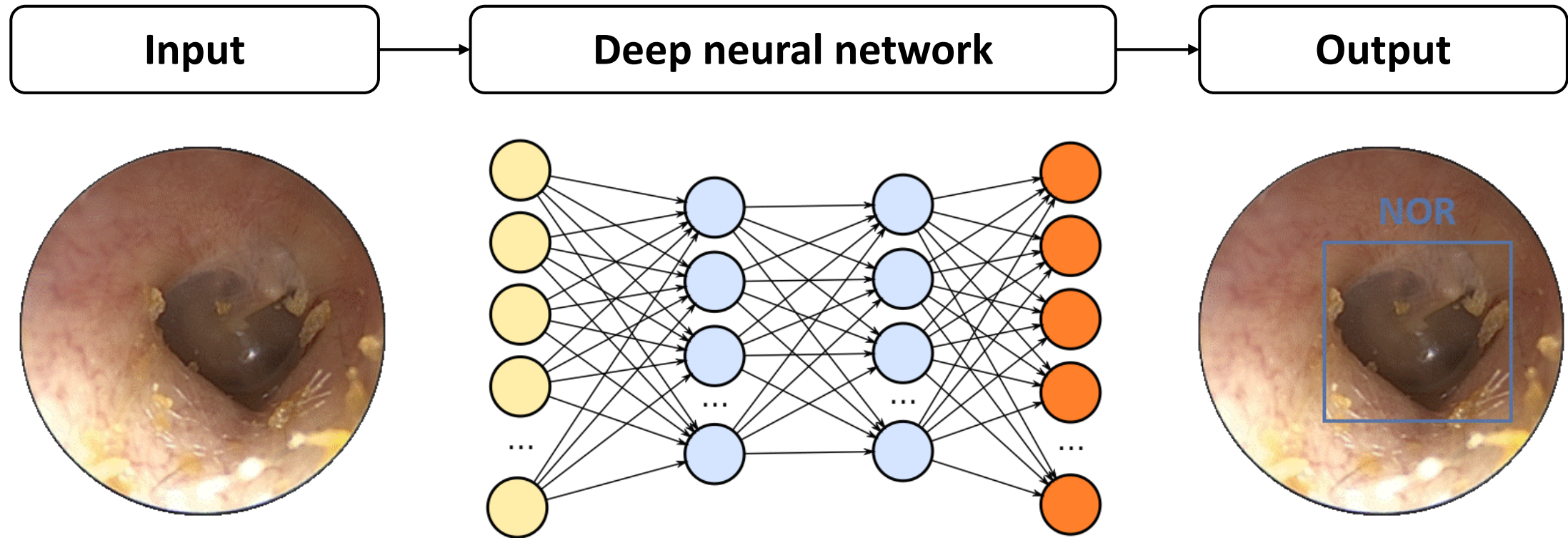


## Visualizing feature latent space using t-SNE [8]



[8] Van der Maaten, Laurens, and Geoffrey Hinton. "Visualizing data using t-SNE." *Journal of machine learning research* 9.11 (2008).

# Model

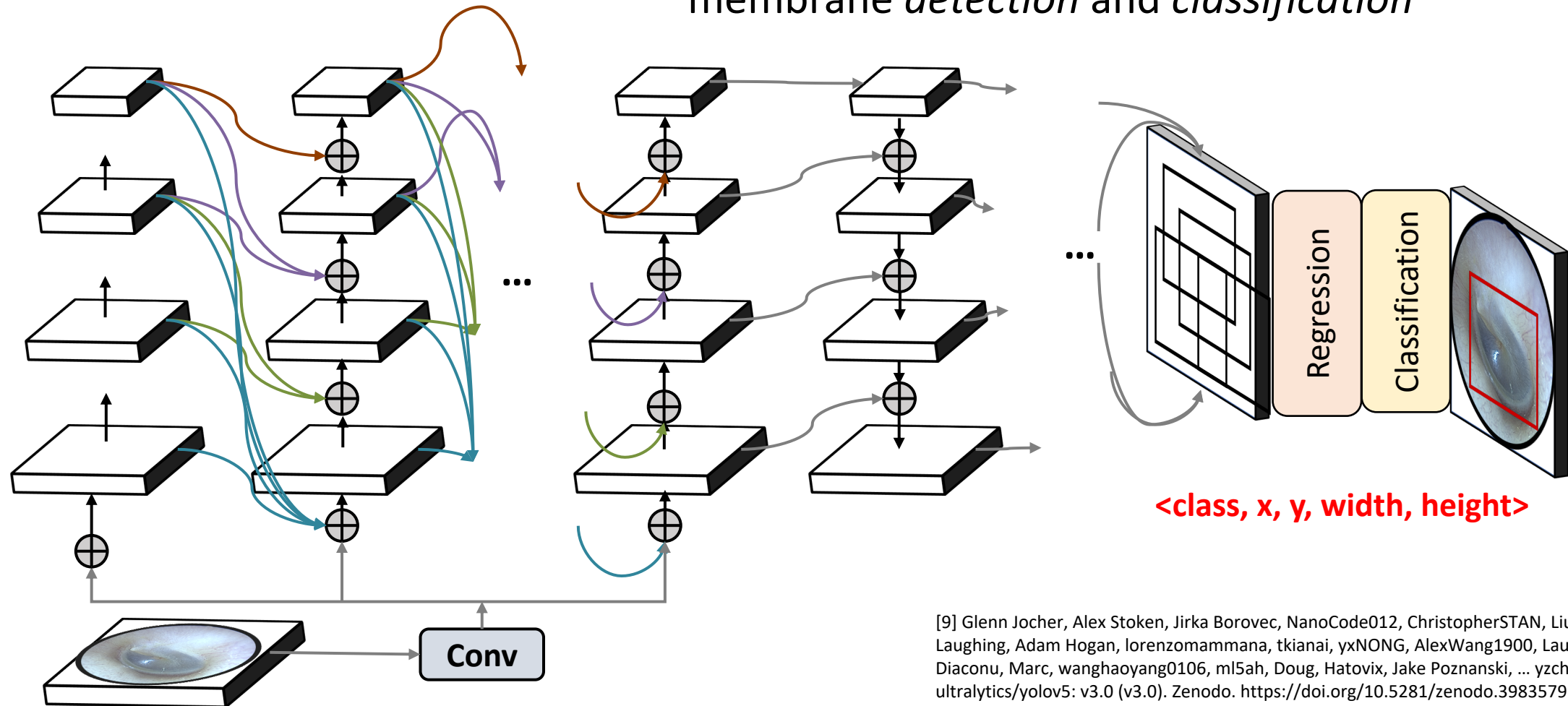


- ✓ We want to model tympanic membrane ROI and its class as output



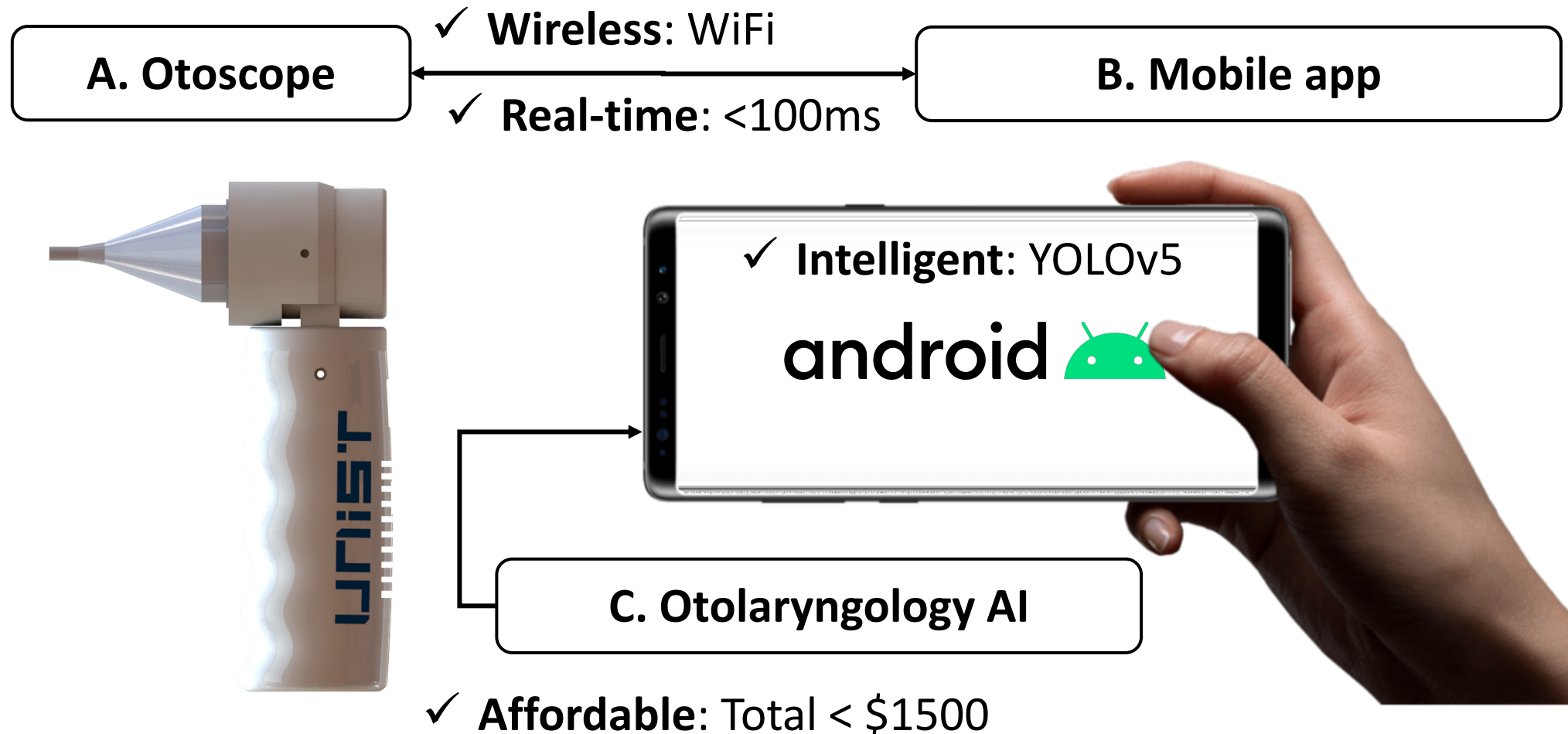
# Model

✓ We utilize YOLOv5 [9] for simultaneous tympanic membrane *detection* and *classification*



[9] Glenn Jocher, Alex Stoken, Jirka Borovec, NanoCode012, ChristopherSTAN, Liu Changyu, Laughing, Adam Hogan, lorenzomamma, tkianai, yxNONG, AlexWang1900, Laurentiu Diaconu, Marc, wanghaoyang0106, ml5ah, Doug, Hatovix, Jake Poznanski, ... yzchen. (2020). ultralytics/yolov5: v3.0 (v3.0). Zenodo. <https://doi.org/10.5281/zenodo.3983579>

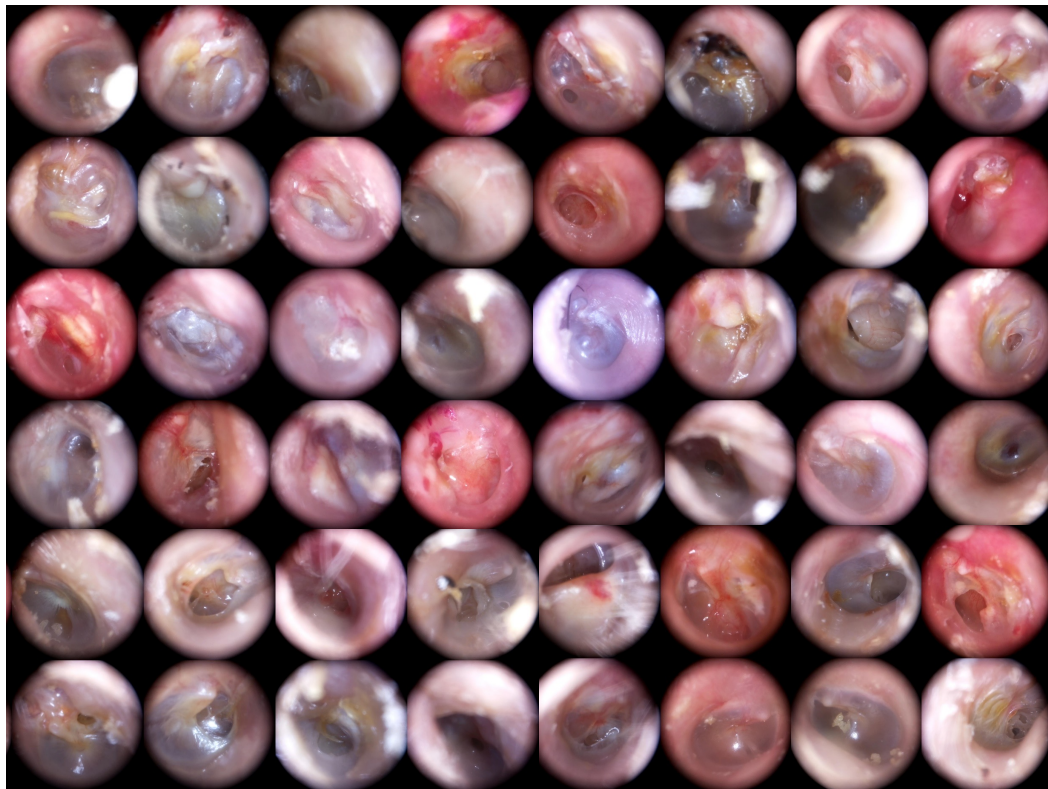
## Ear examination platform



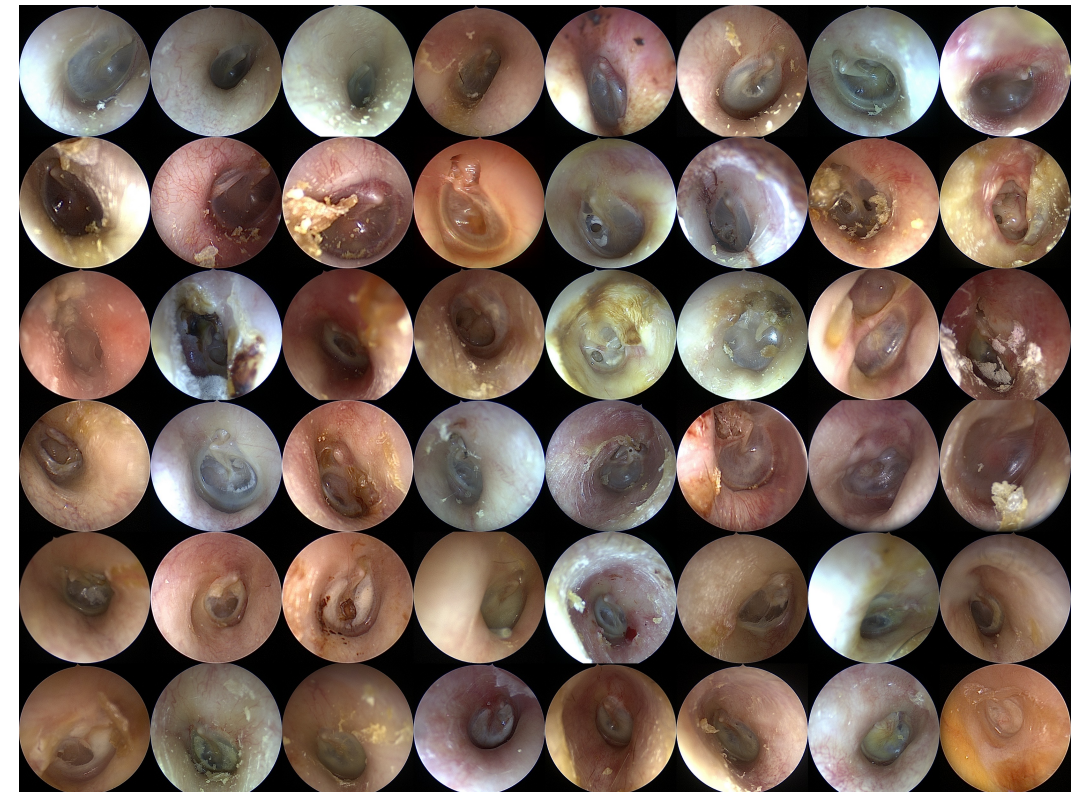
# 3. Validation

# Image quality

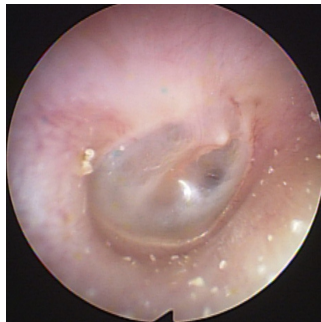
Ours



Clinical



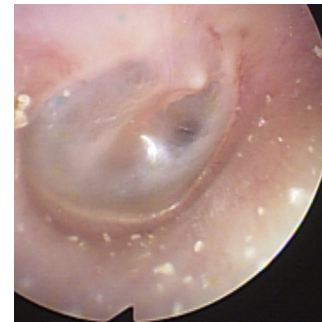
# Image augmentation



Original  
image



**Blurring**



75% random  
crop



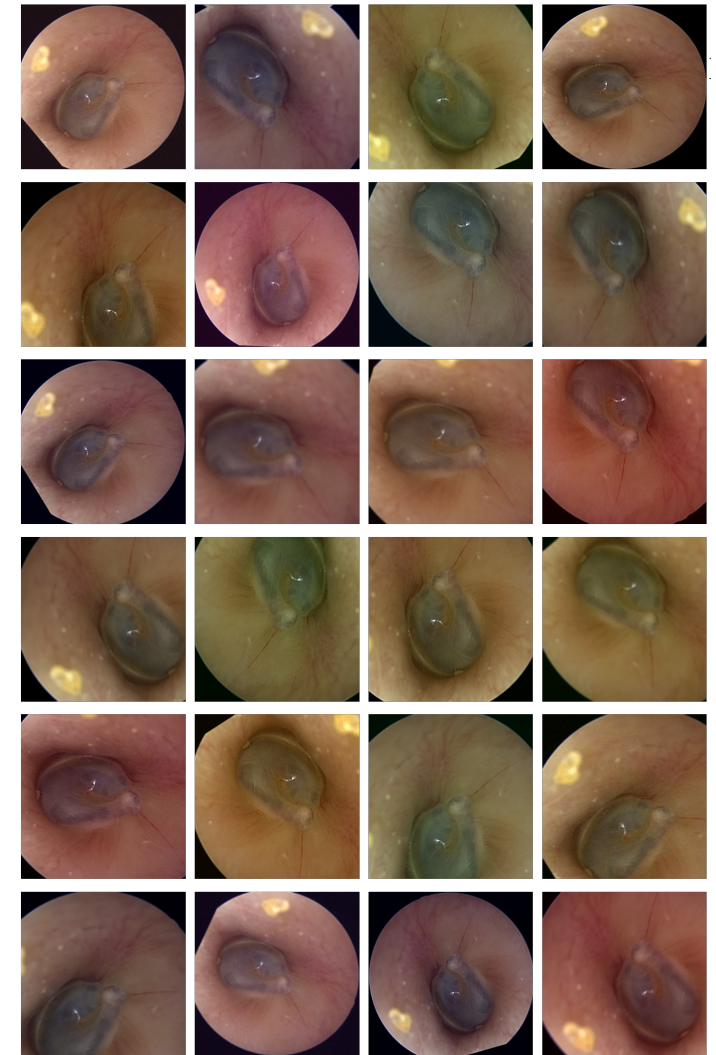
Random  
brightness



Random  
contrast

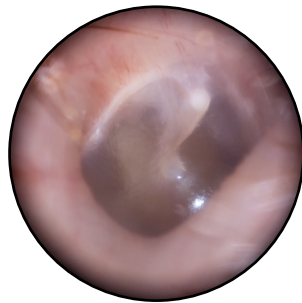


**Random  
color shift**

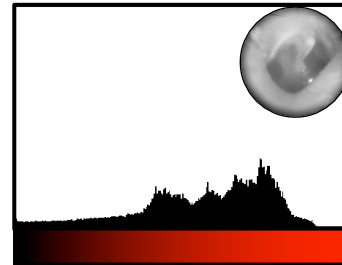
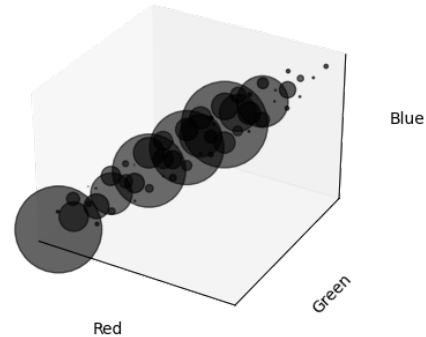


# Image color

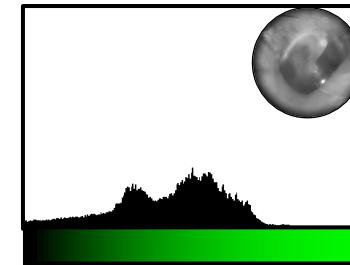
Ours



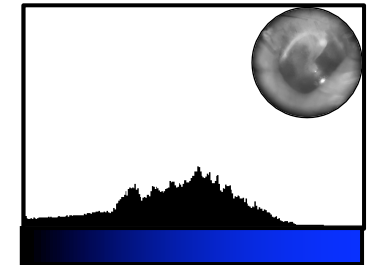
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$\mu = 121.7$   
 $\sigma = 67.9$

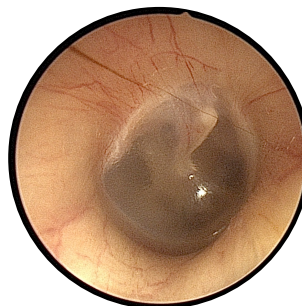


$\mu = 93.3$   
 $\sigma = 55.9$

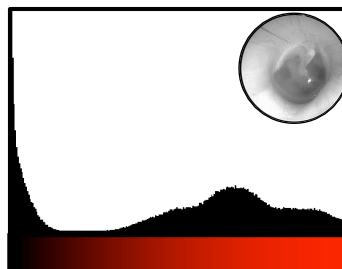
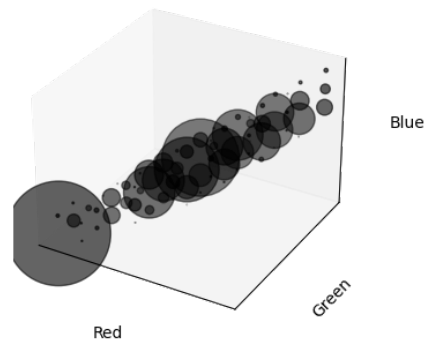


$\mu = 95.3$   
 $\sigma = 56.9$

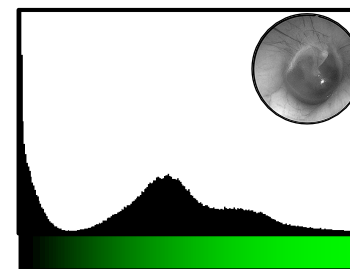
Clinical



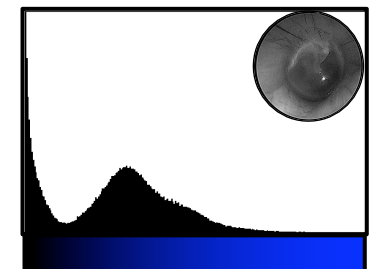
(512,512,3)



$\mu = 129.2$   
 $\sigma = 81.6$



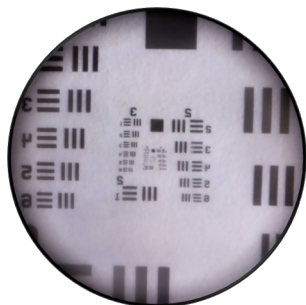
$\mu = 93.2$   
 $\sigma = 62.1$



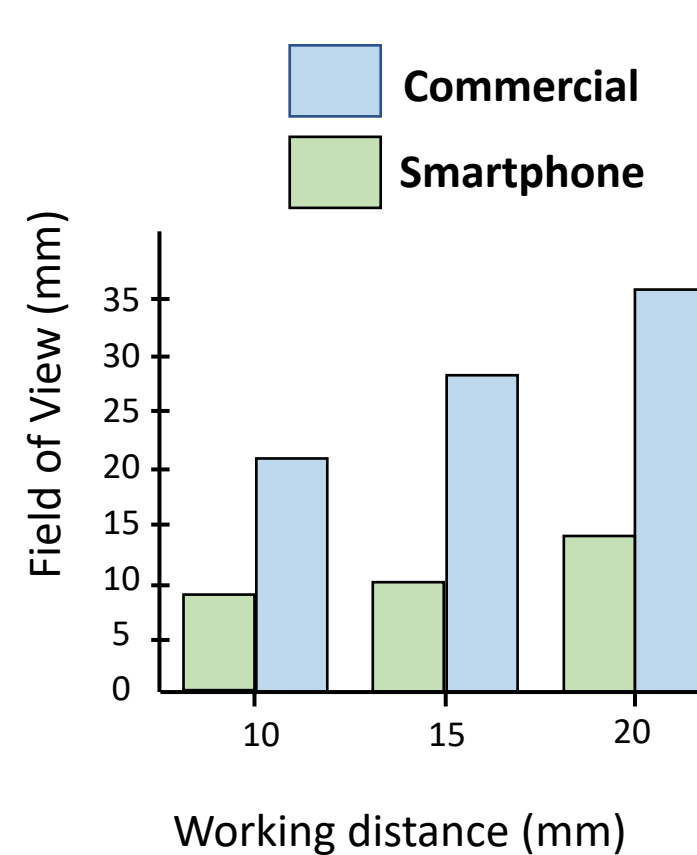
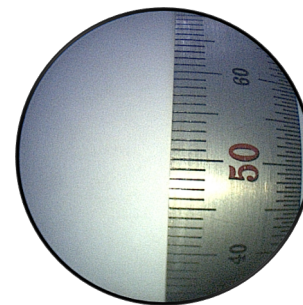
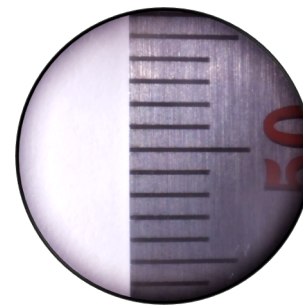
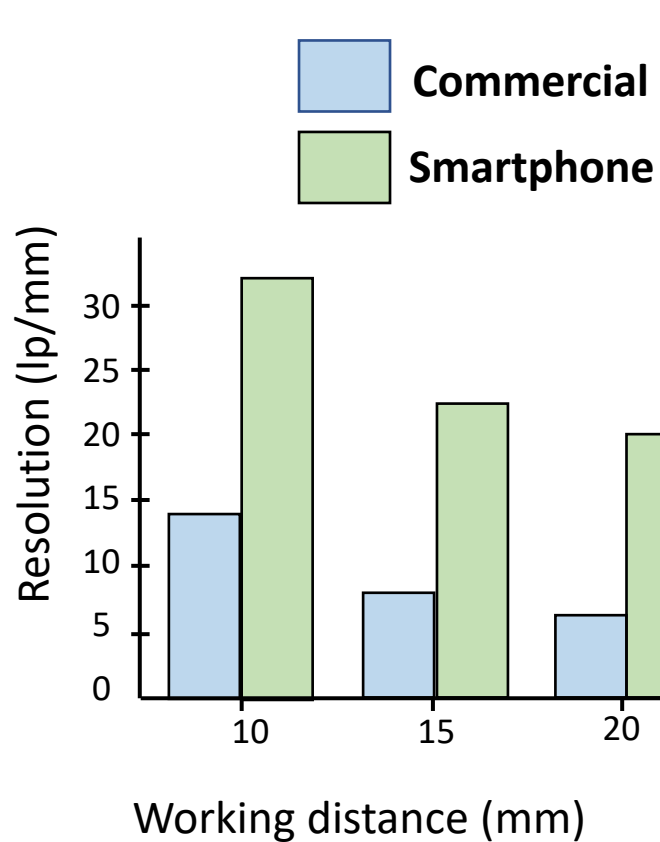
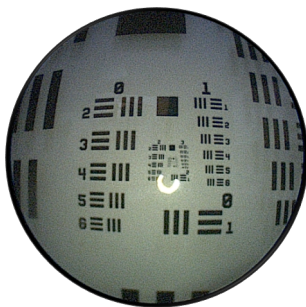
$\mu = 65.7$   
 $\sigma = 45.2$

# Image quality

Ours

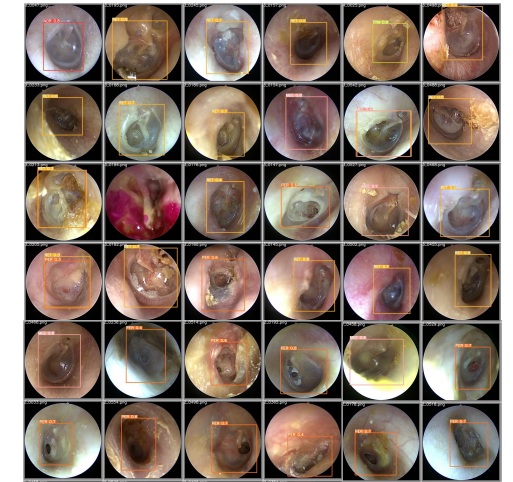
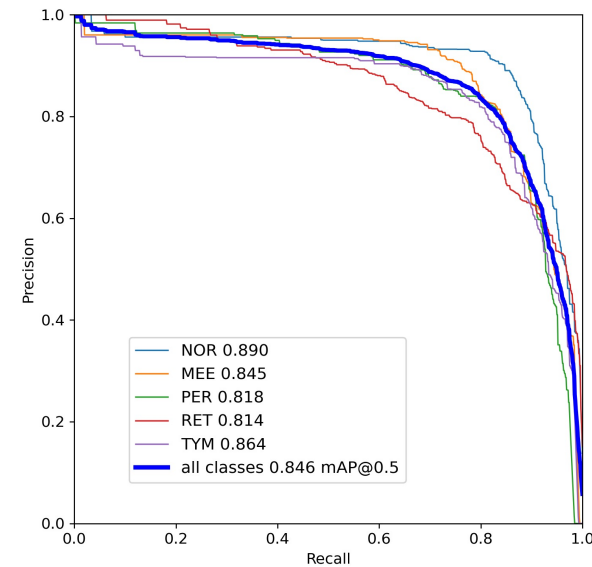
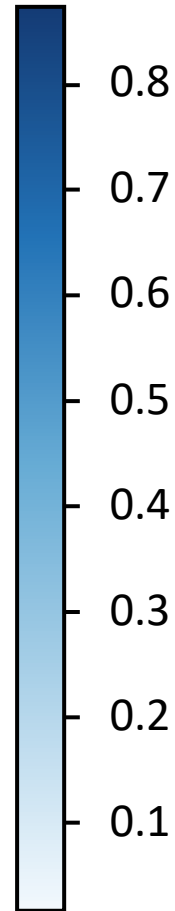


Clinical



# AI performance

Predicted	NOR	0.86	0.02	0.01	0.01	0.06
	MEE	0.04	0.81		0.06	0.03
	PER	0.01	0.02	0.79	0.07	0.02
	RET	0.03	0.08	0.06	0.75	0.04
	TYM	0.03	0.03	0.05	0.04	0.79
		NOR	MEE	PER	RET	TYM
	True					



## Evaluation

Metrics	Value
Epochs	300
Precision	0.834
Recall	0.808
mAP <sub>0.5</sub>	0.87
mAP <sub>0.5:0.95</sub>	0.568

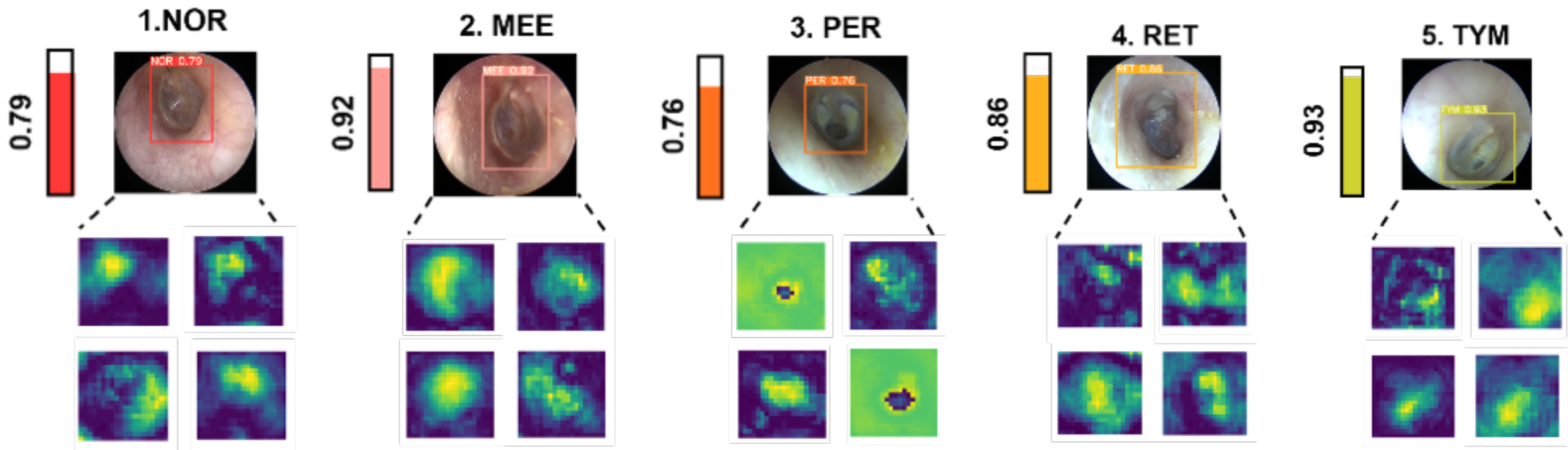
## Configuration

Parameter	Description
Layers	224
Hyperparams	7,064,698
GFLOPs	16.4
Size (MB)	14.4
Fp16 (MB)	6.9



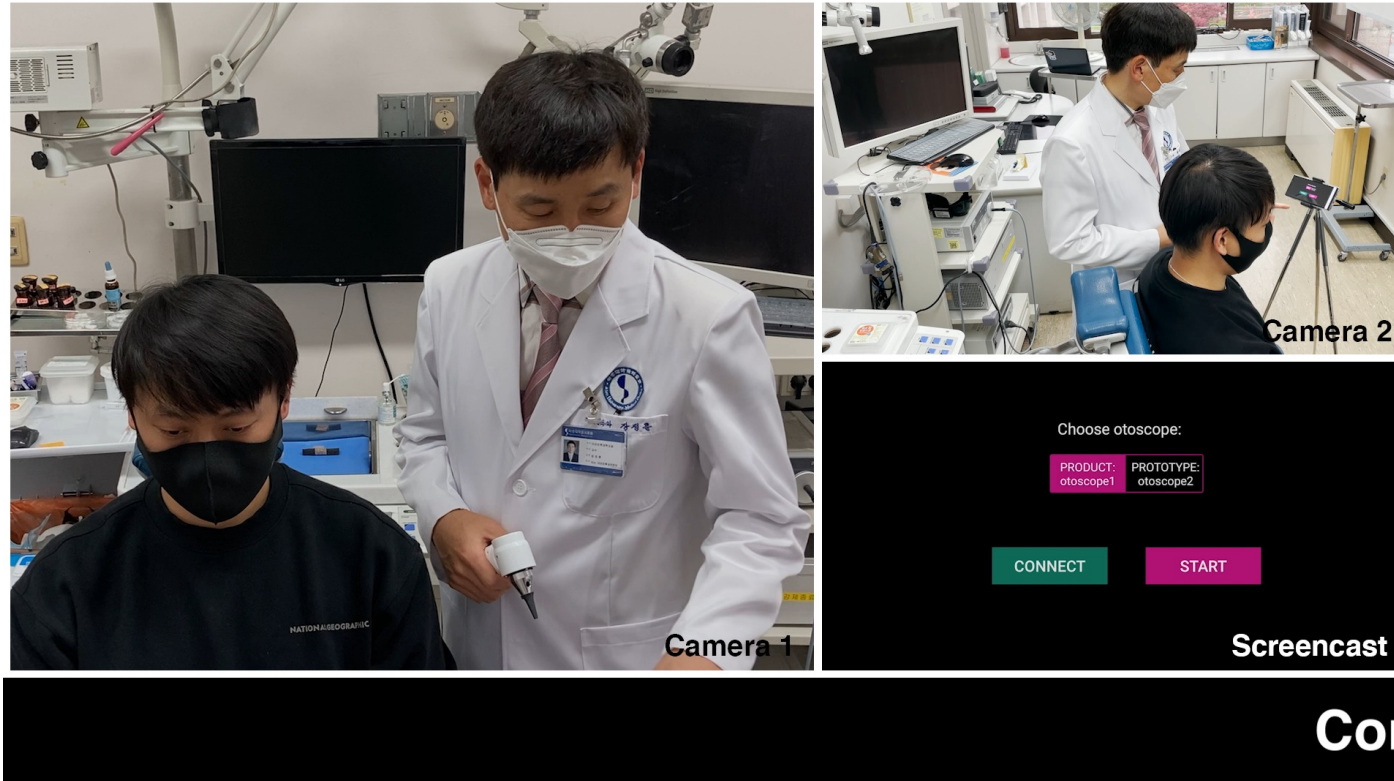
# AI performance

## Class probability identification in validation set



## Reasoning results with class activation maps

# Demo



(In collaboration with otologist (JHJ) from Ajou University Hospital, Suwon, South Korea, 2021)

# 4. Conclusion

# Conclusion

- ✓ **Specialist ear examination remains common issue in low-income countries**

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- ✓ **Specialist ear examination remains common issue in low-income countries**
- ✓ **To overcome, procedure needs to assistive and affordable technology**

# Conclusion

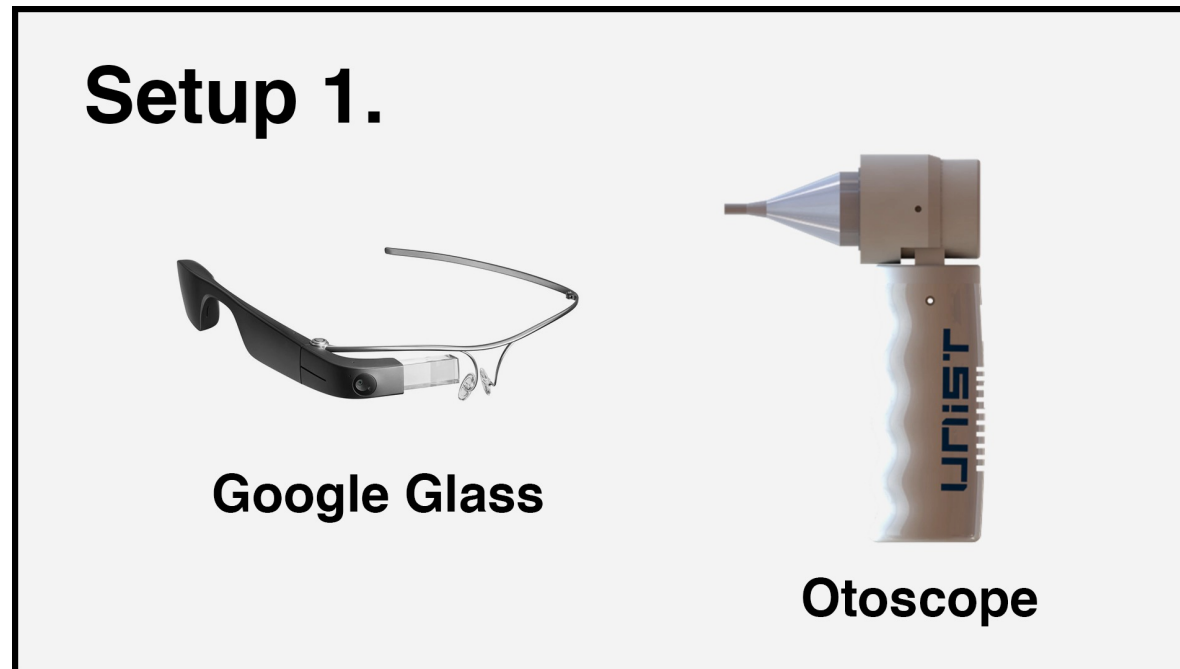
- ✓ **Specialist ear examination remains common issue in low-income countries**
- ✓ **To overcome, procedure needs assistive and affordable technology**
- ✓ **Here, we propose a mobile, deep learning-assisted otoscope**

# Conclusion

- ✓ **Specialist ear examination remains common issue in low-income countries**
- ✓ **To overcome, procedure needs assistive and affordable technology**
- ✓ **Here, we propose a mobile, deep learning-assisted otoscope**
- ✓ **Our results demonstrated high diagnostic accuracy indicating potential to become a viable screening solution in low-resource, non-specialist settings.**

# Perspectives

- ✓ Telemedicine applications with augmented reality





# Contributors

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**Dr. Joon S. Yoo**

Incipian LLC

**Professor Gil-Jin Jang**

Kyungpook National University

**Professor Jeong Hun Jang**

Ajou University Hospital

**Professor Woonggyu Jung**

UNIST



Incipian



# Acknowledgement



- ✓ This research was supported by a grant of the Korea Health Technology R&D Project through the Korea Health Industry Development Institute (KHIDI), funded by the Ministry of Health & Welfare, Republic of Korea (grant number: HP20C0032).
- ✓ This work was supported by the Korea Medical Device Development Fund grant funded by the Korea government (the Ministry of Science and ICT, the Ministry of Trade, Industry and Energy, the Ministry of Health & Welfare, the Ministry of Food and Drug Safety) (Project Number: 1711137934, KMDF\_PR\_20200901\_0024-03).

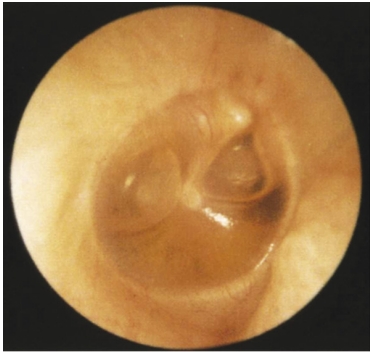
# Thank you for attention!



# Q&A

Correspondence:  
[s.askaruly@gmail.com](mailto:s.askaruly@gmail.com)

### Middle ear effusion



Thick or sticky fluid behind the eardrum in the middle ear. [1]

### Perforation



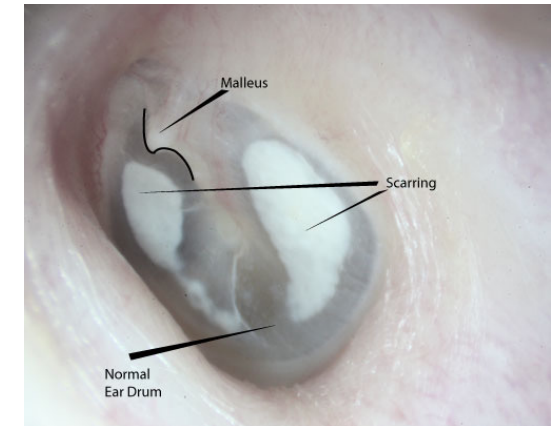
A ruptured eardrum (tympanic membrane perforation) is a hole or tear in the thin tissue that separates your ear canal from your middle ear (eardrum). [2]

### Retraction



Is a condition where the tympanic membrane, or eardrum, gets pulled toward the middle of ear. [3]

### Tympanosclerosis



Tympanosclerosis is the medical term for scarring of the ear drum. Scarring occurs after the ear drum is injured or after surgery. Commonly a small white area can be seen after a person has had middle ear ventilation tubes. [4]

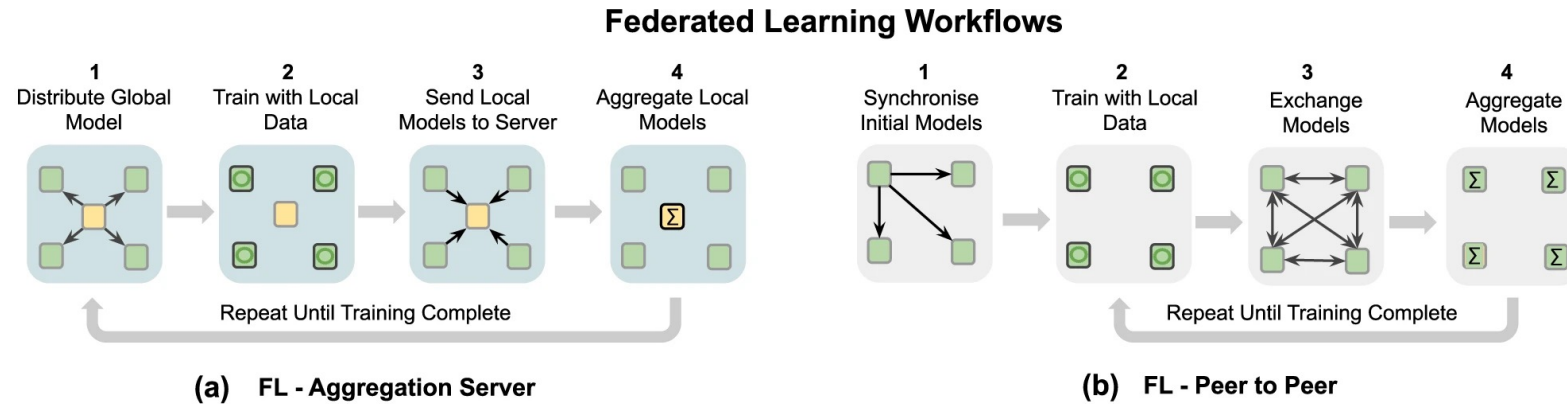
[1] <https://medlineplus.gov/ency/article/007010.htm>

[2] <https://www.mayoclinic.org/diseases-conditions/ruptured-eardrum/symptoms-causes/syc-20351879>

[3] <https://www.webmd.com/a-to-z-guides/what-is-a-tympanic-membrane-retraction>

[4] <https://med.uth.edu/orl/online-ear-disease-photo-book/chapter-15-miscellaneous/tympanosclerosis/>

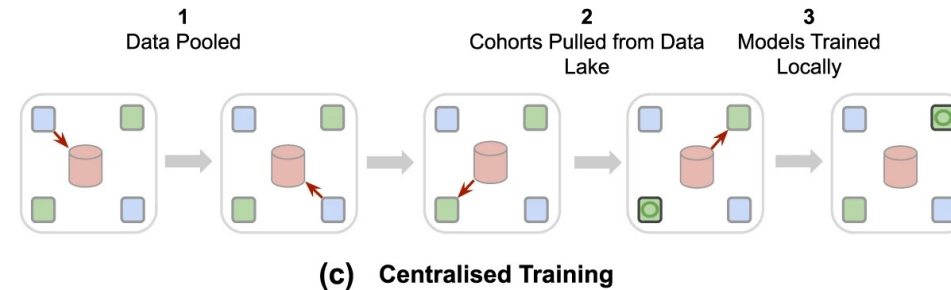
# Federated Learning



**Key**

- Aggregation Server
- Training Node
- $\Sigma$  Model Aggregation
- $\leftarrow$  Weight/Gradient Exchange
- Central Data Lake
- Data Donor
- $\rightarrow$  Time
- ↔ Medical Data Exchange
- Local Training

**Centralised Data Lake**



Rieke, Nicola, et al. "The future of digital health with federated learning." *NPJ digital medicine* 3.1 (2020): 1-7.