





# Deep Neural Networks for Noise Reduction under Hearing Aid Side Conditions

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IHCON 2018, Lake Tahoe, CA, United States

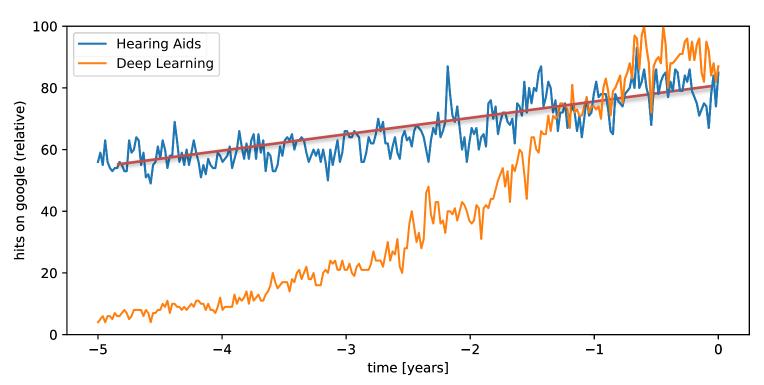








#### **Attention**



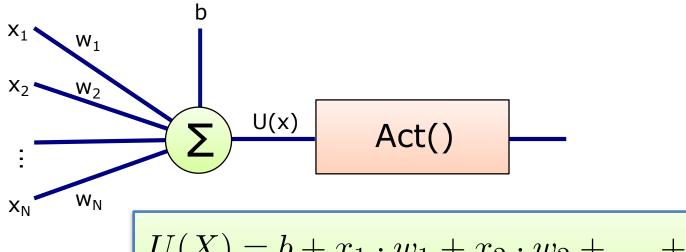
[google trends: "hearing aids" vs. "deep learning"]







#### What is deep learning anyway?



 $U(X) = b + x_1 \cdot w_1 + x_2 \cdot w_2 + \ldots + x_N \cdot w_N$  Linear Regression

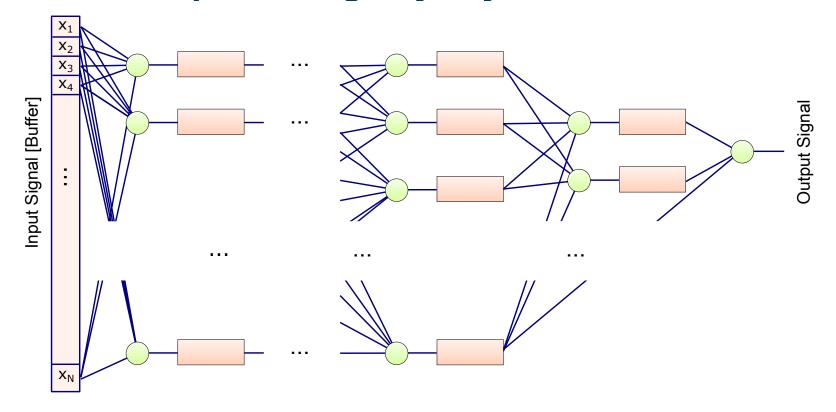
**Nonlinearity** 







## What is deep learning anyway?

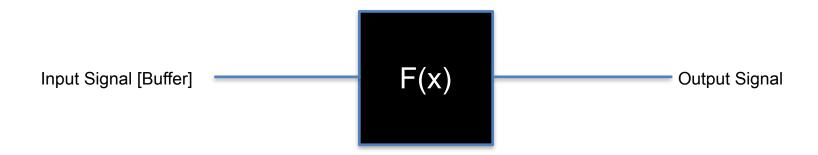








## What is deep learning anyway?



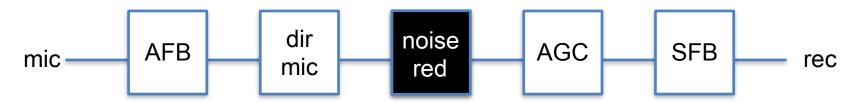
- Not very smart, but very effective.
- Contents of the black box may seem unclear.
- Problem for medical devices.







## **Learning known Operators [MS17]**



- We can exchange a single operator in the chain by a neural network.
- Restrictions to the structure of output and network help constrain the problem to be solved.

[MS17]: Maier et al., 2017: arxiv:1712.00374







## **Hearing Aid Side-Conditions**

Limited processing power

Robustness required in any environment

Group delay > 10 ms objectionable to hearing aid wearers [AT00]

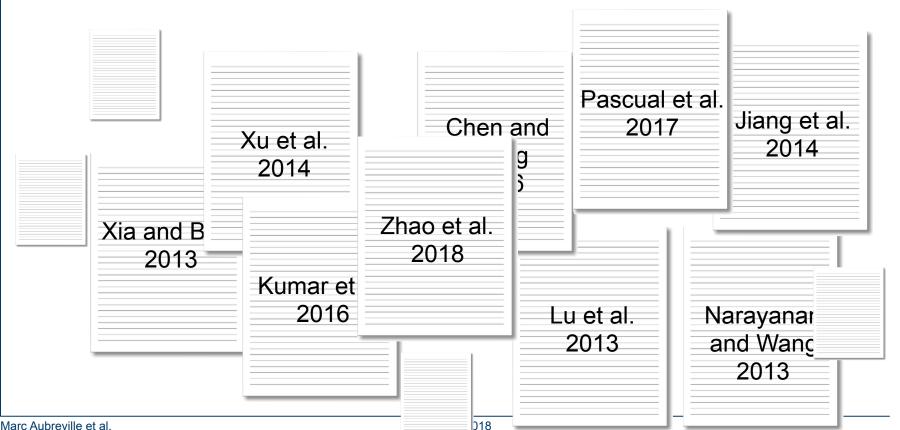
[AT00]: J. Agnew and J. M. Thornton, JAAA (2000) 11:330-360







## State of the (deep) art

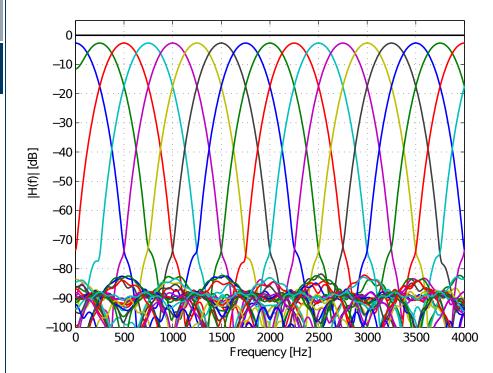








## **Hearing Aid Filterbank**



[BS08]: R. Bäuml and W. Sörgel, EUSIPCO 2008

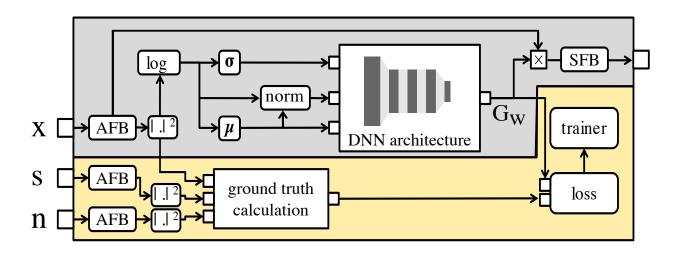
- State-of-the-art hearing aid filter bank
- 48 channels, uniform
- group delay: ~ 6 ms







#### **Architecture**



- Input: Log-levels, normalization on window scope
- DNN with 3 hidden layers, 2048 nodes each.

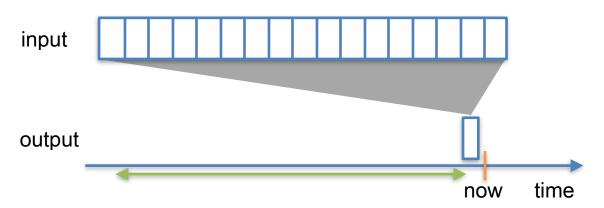






## **Temporal context**

- Temporal context should be 200-300ms, due to structure of speech (~ 4 Hz syllable rate). [HS77]
- Symmetrical context leads to high latency
- Proposal: Asymmetrical context with rich past knowledge



[HS77]: Houtgast and Steeneken, JASA 1985:77(3)







## Our training database

- 49 real-world noise recordings
  - recorded with hearing aid microphones
  - Mixed to achieve multi-noise conditions (Kumar et al.)







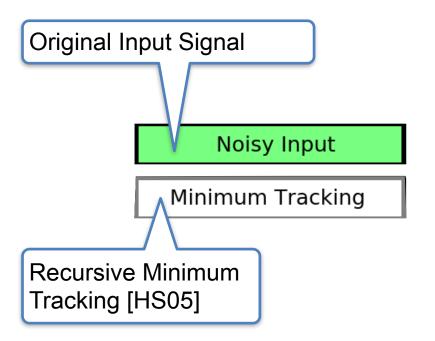
- 260 clean speech signals (EUROM, german sentences)
- Train/test split on source signal level.

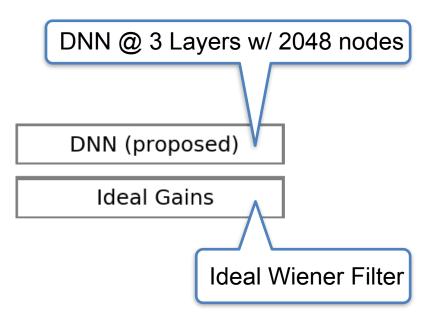






#### **Results - Conditions**



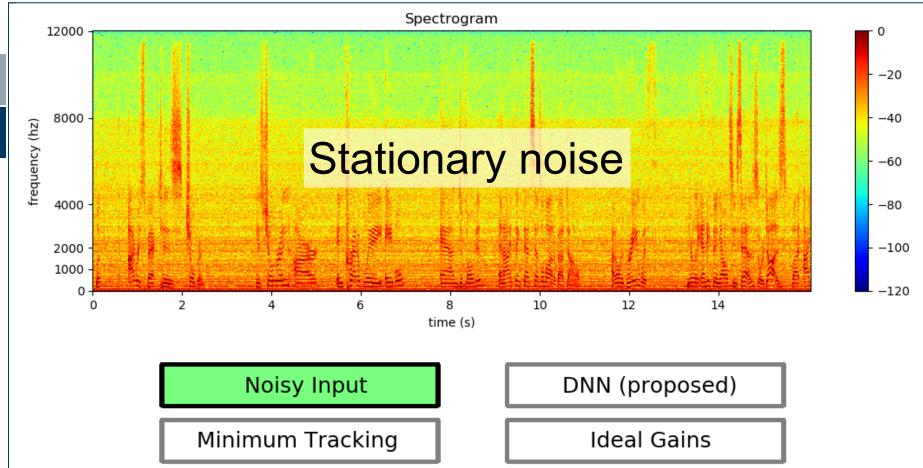


[HS05]: E. Hänsler, G. Schmidt, Wiley&Sons, 2005





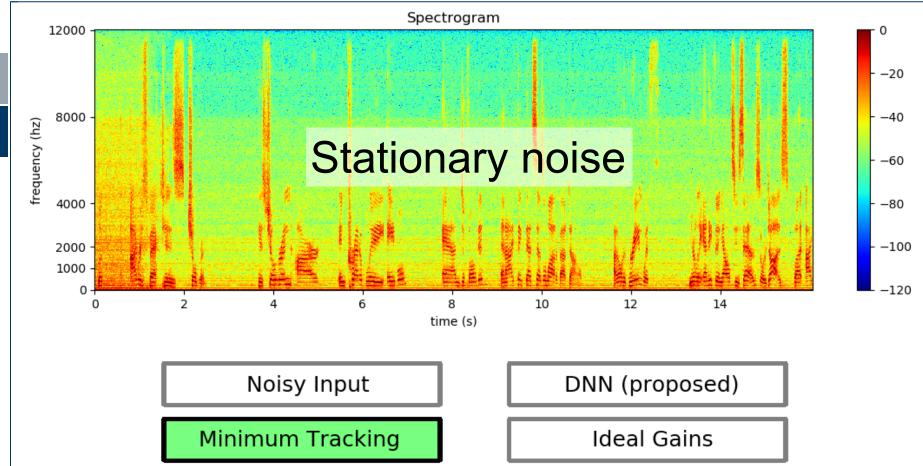








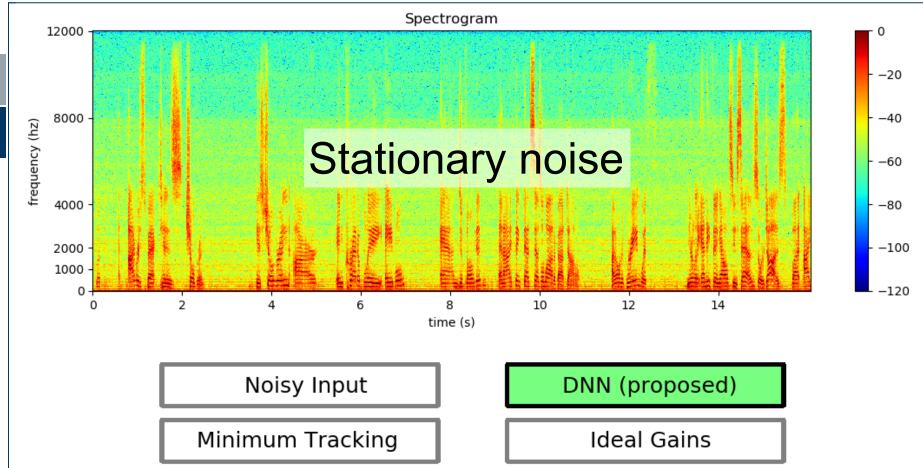








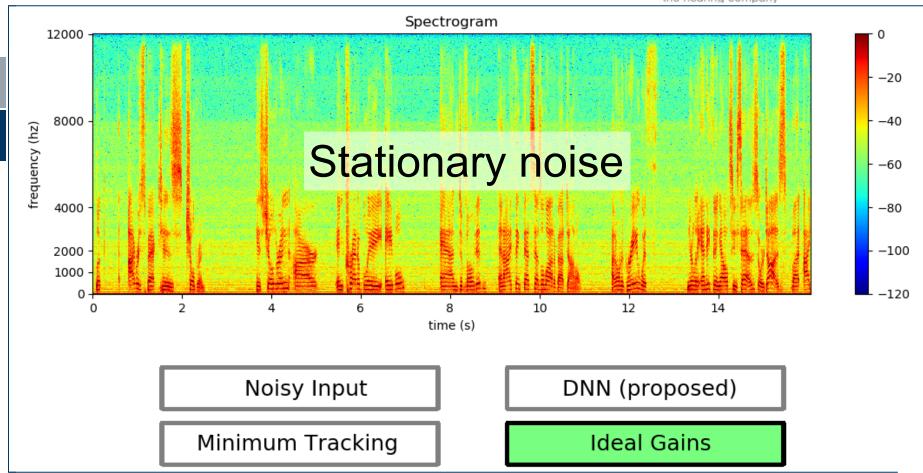








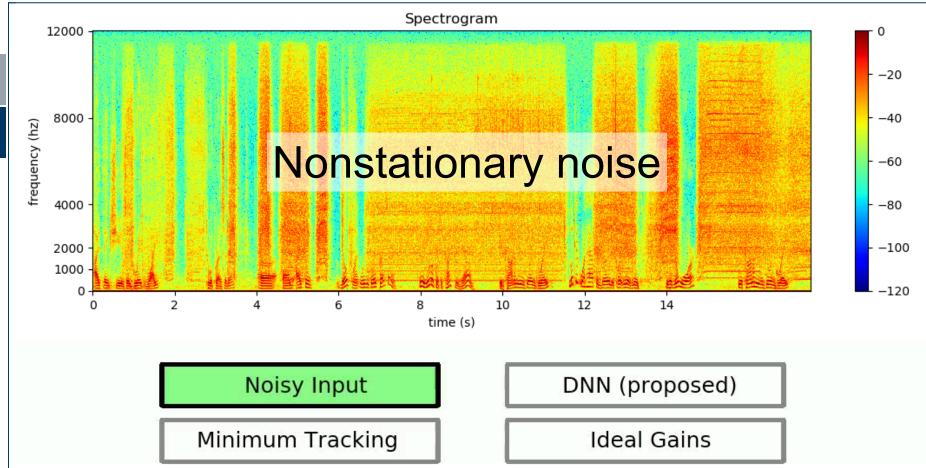










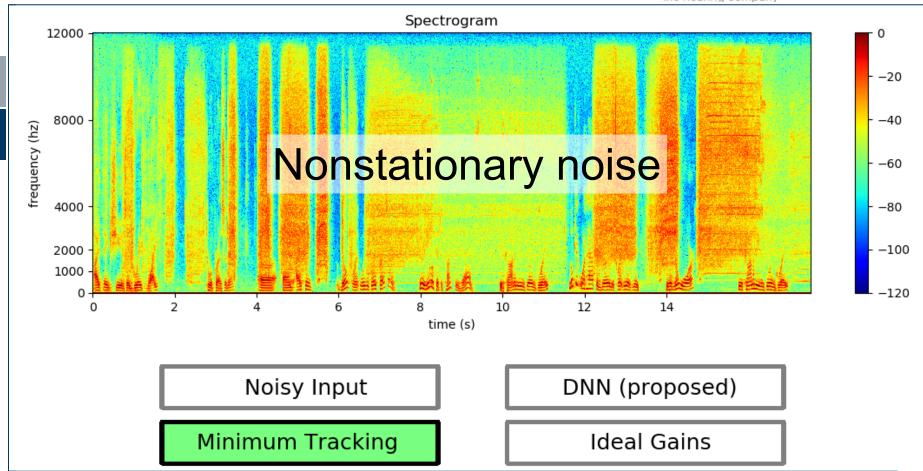


Marc Aubreville et al. IHCON 2018 2018-08-18





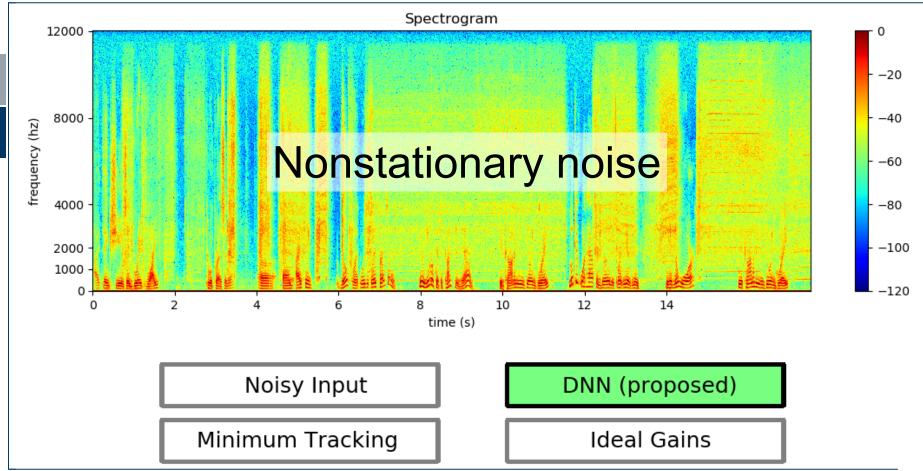








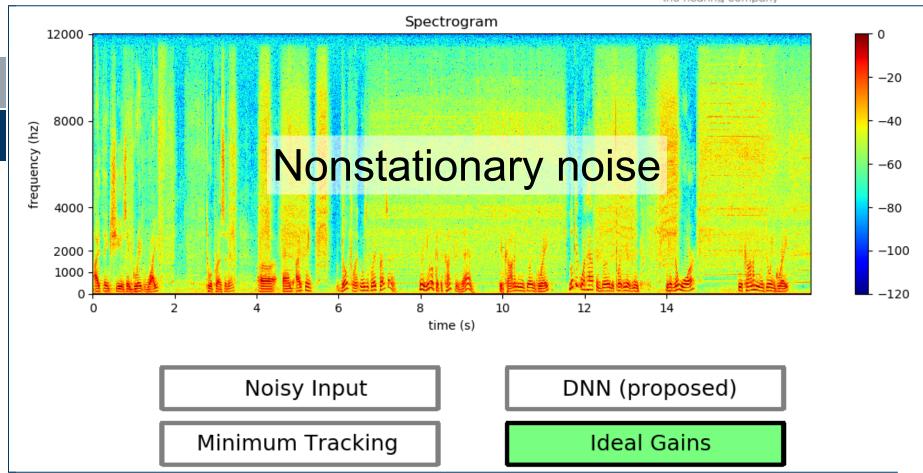














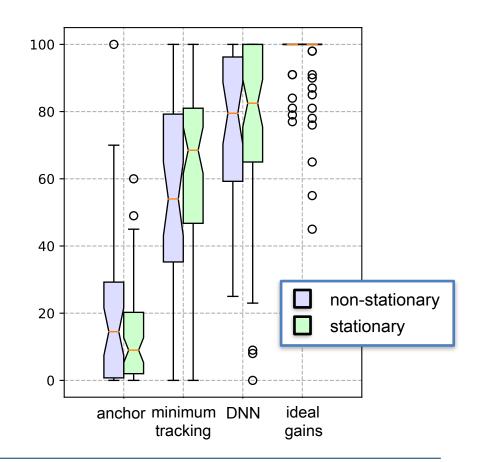




#### **Subjective Results**

(MUSHRA, N=20)

- DNN as superior to recursive minimum tracking baseline
- Benefits especially for nonstationary signals.
- Still imperfect quality for low SNRs (phase distortion)



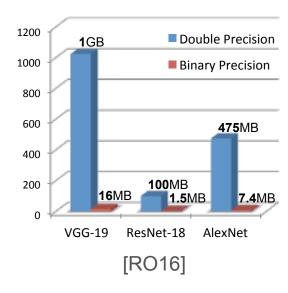






# Deep Learning is too complex for hearing aids?

- DNNs are not dependent on floating point units.
- Even binary operations might be sufficient. (XNOR-Net [RO16])
- Potential for revival of analog computing [LY15]



[RO16] M. Rastegari et al., 2016, arxiv: 1603.05279 [LY15] Y. Lu et al, 2015, IEEE J. Solid State Circuits 50(1)







## **Summary: Hearing Aid Side-Conditions**

- Limited processing power
  - Likely not unsolvable in the future.
- Robustness required in any environment
  - Promising, needs more evaluation.
- Group delay > 10 ms objectionable to hearing aid wearers [AT00]
  - Total group delay = 8 ms.

[AT00]: J. Agnew and J. M. Thornton, JAAA (2000) 11:330-360







#### Thank you.

For more details:
IWAENC 2018 paper
<a href="https://arxiv.org/abs/1805.01198">https://arxiv.org/abs/1805.01198</a>









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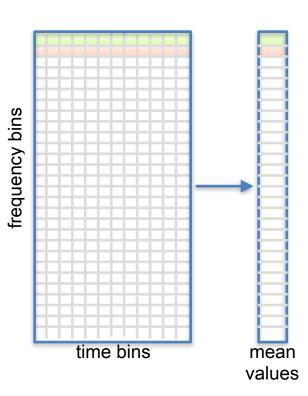
#### **Normalization**

Normalization across frequency:

$$X_{\text{norm}}(k,f) = X(k,f) - \frac{1}{\tau_1 + 1 + \tau_2} \sum_{k=-\tau_1}^{\tau_2} X(k,f)$$

X(k, f): Filterbank levels in log-scale

- Completely level independent
- Level information is fed as side-information to DNN

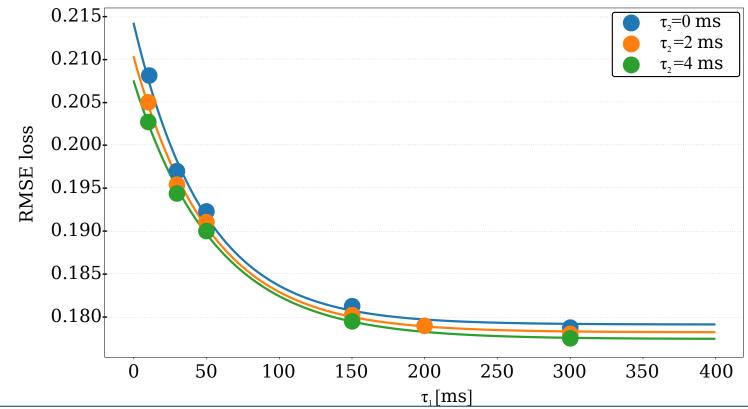








#### How much context is needed?









#### **Context Size Influence on Gain Prediction**

**Example: Fricative** 

