

The Sensitivity Matrix

Integrating Perception into the Flexcode Project

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Flexcode Seminar Lannion

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FlexCode

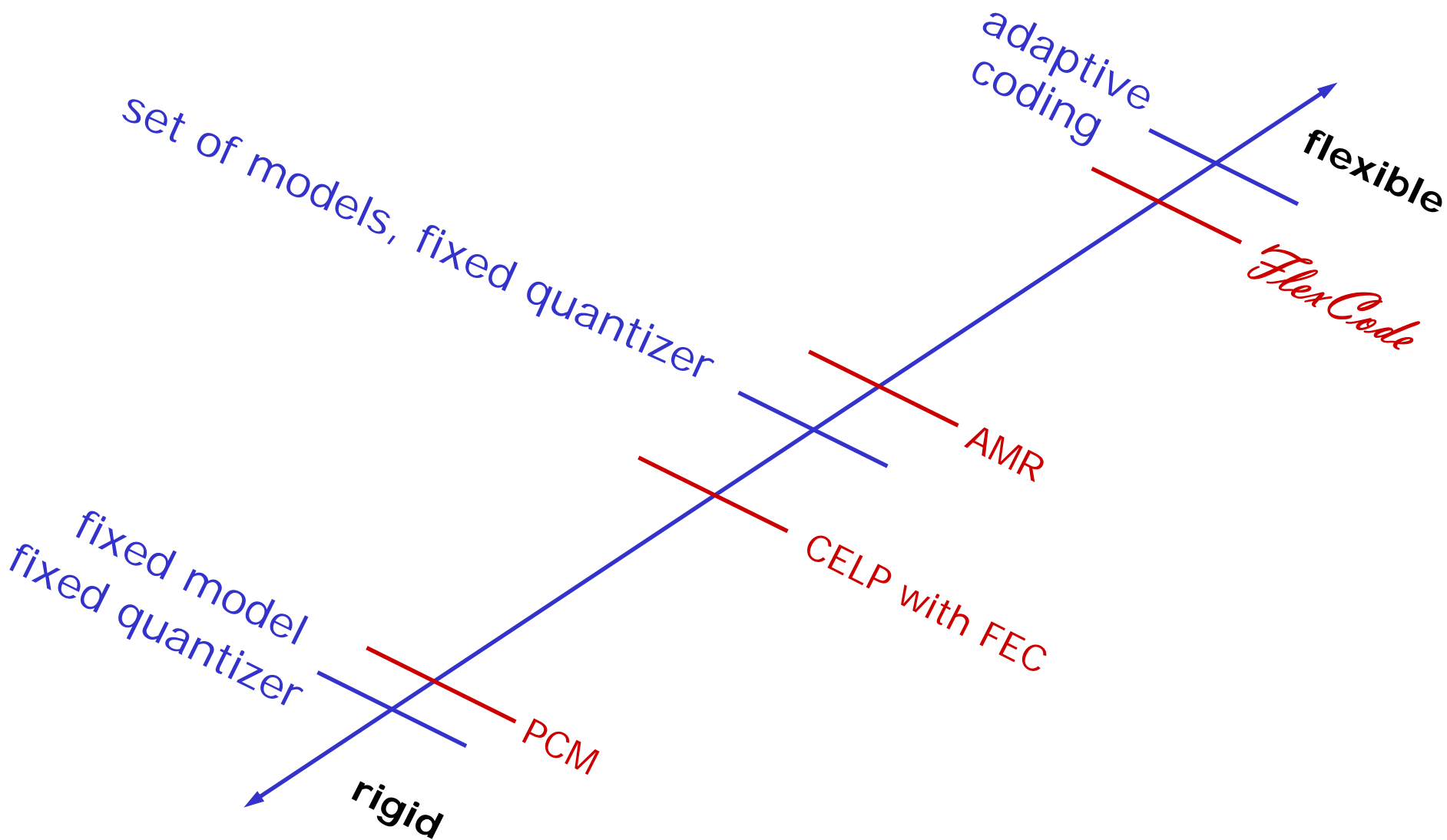
- Flexcode in a Nutshell
- The Sensitivity Matrix
- Coding with the Sensitivity Matrix
- Open Issues

Who?



- Heterogeneity of networks increasing
- Networks inherently variable (mobile users)

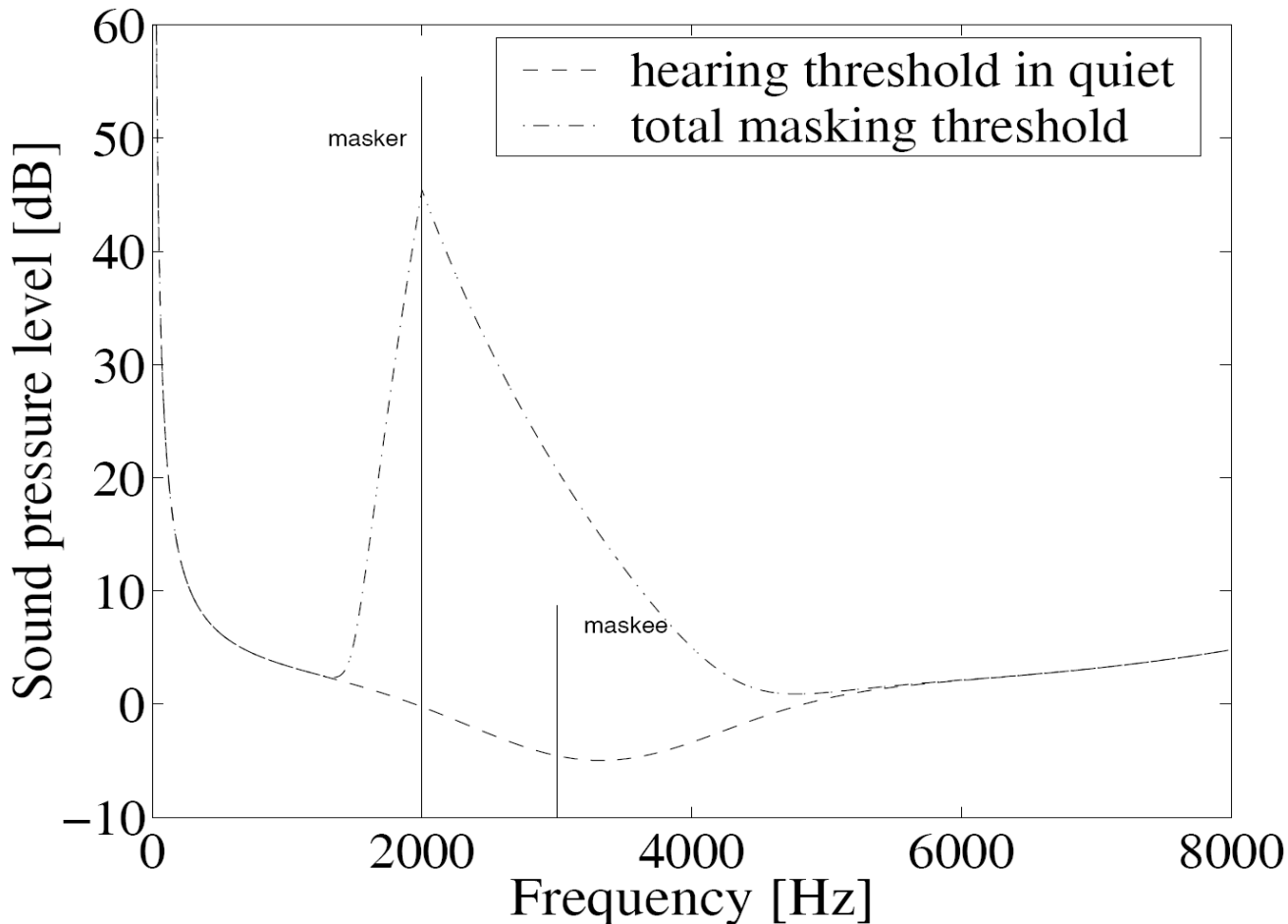
- **But:**
 - Coders not designed for specific environment
 - Coders inflexible (codebooks and FEC)
 - Feedback channel underutilized



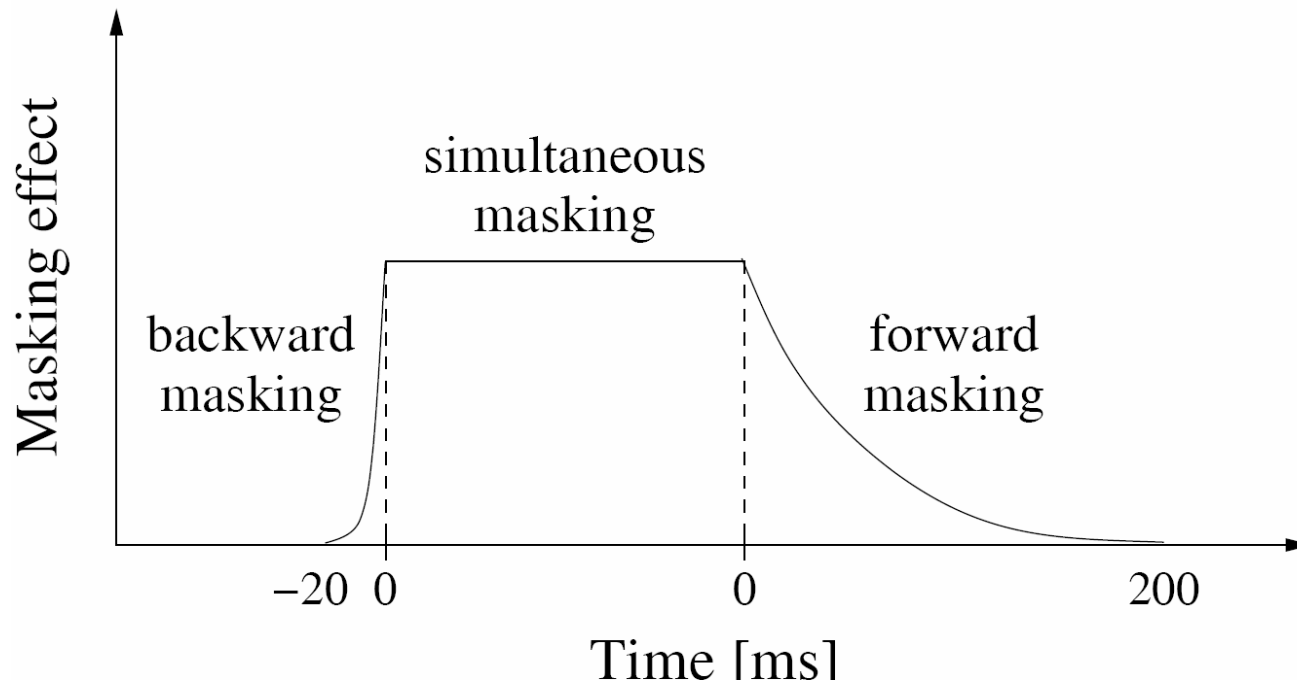
- Tools include
 - Models of source, channel, **receiver**
 - High-rate quantization theory
 - Multiple description coding (MDC)
 - Iterative source-channel decoding
 - Distortion measures using the **sensitivity matrix**

- Flexcode in a Nutshell
- The Sensitivity Matrix
 - Why, what and how?
 - What can it tell us?
- Coding with the Sensitivity Matrix
- Open Issues

- Simultaneous masking (frequency masking)

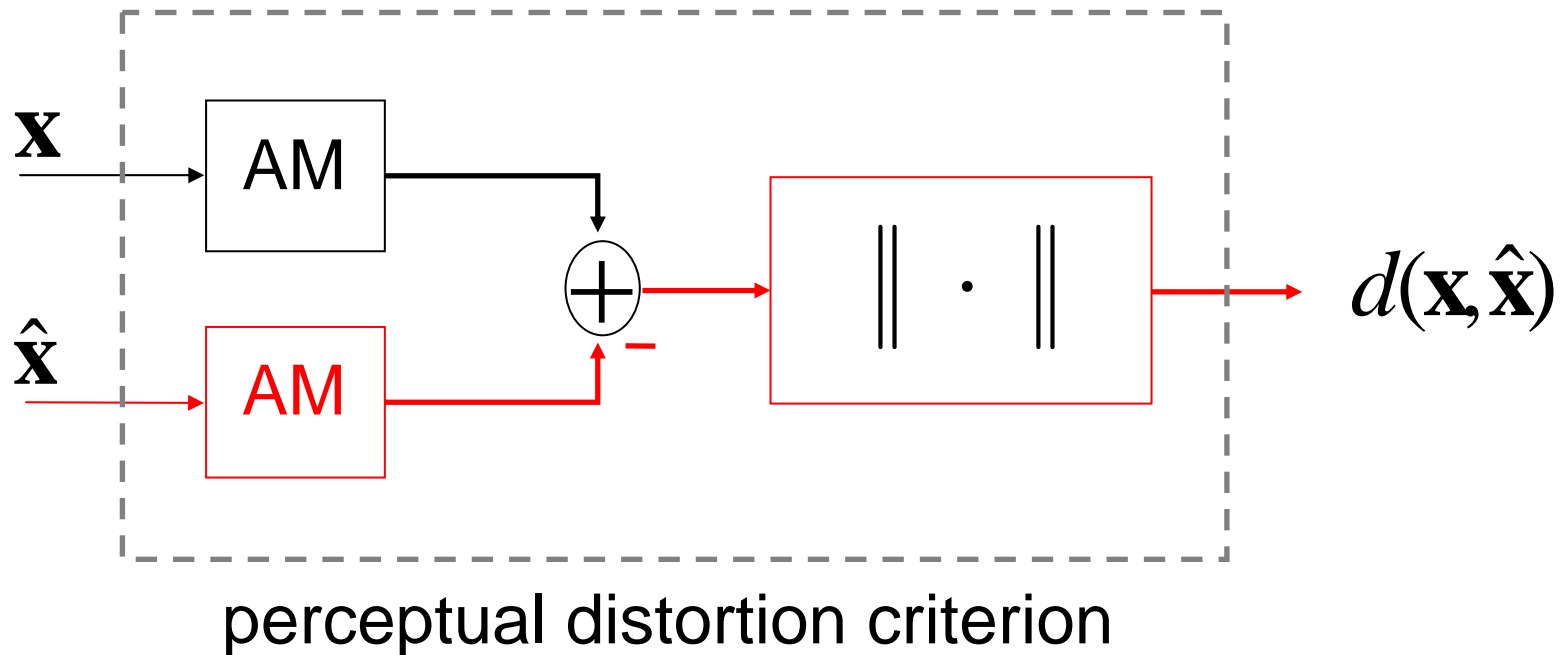


- Non-simultaneous masking (time masking)



- Not accounted for by simple auditory models
- Coding standards use (heuristic) work-arounds

- Merge the worlds of **auditory modeling** and **audio coding**
- Advanced auditory models **too complex** for coding



- Sensitivity Matrix **M** from Taylor expansion

$$d(\mathbf{x}, \hat{\mathbf{x}}) \approx \frac{1}{2} (\mathbf{x} - \hat{\mathbf{x}})^H \mathbf{M}_x(\mathbf{x}) (\mathbf{x} - \hat{\mathbf{x}}), \quad d(\mathbf{x}, \hat{\mathbf{x}}) \rightarrow 0$$

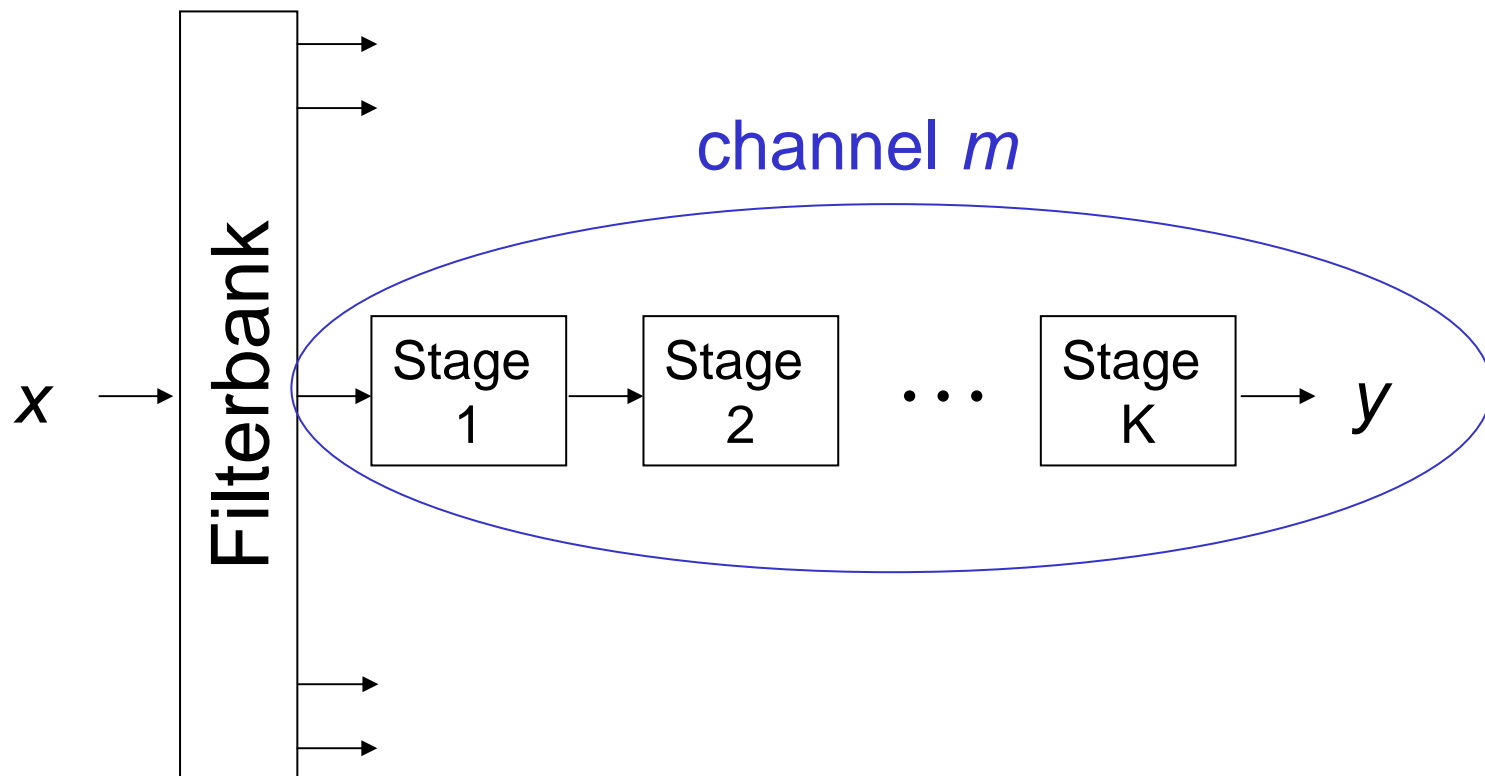
$$[\mathbf{M}_x(\mathbf{x})]_{i,j} = \left. \frac{\partial^2 d(\mathbf{x}, \hat{\mathbf{x}})}{\partial \hat{x}_i \partial \hat{x}_j} \right|_{\hat{\mathbf{x}}=\mathbf{x}}$$

➤ **Complexity** problem solved!

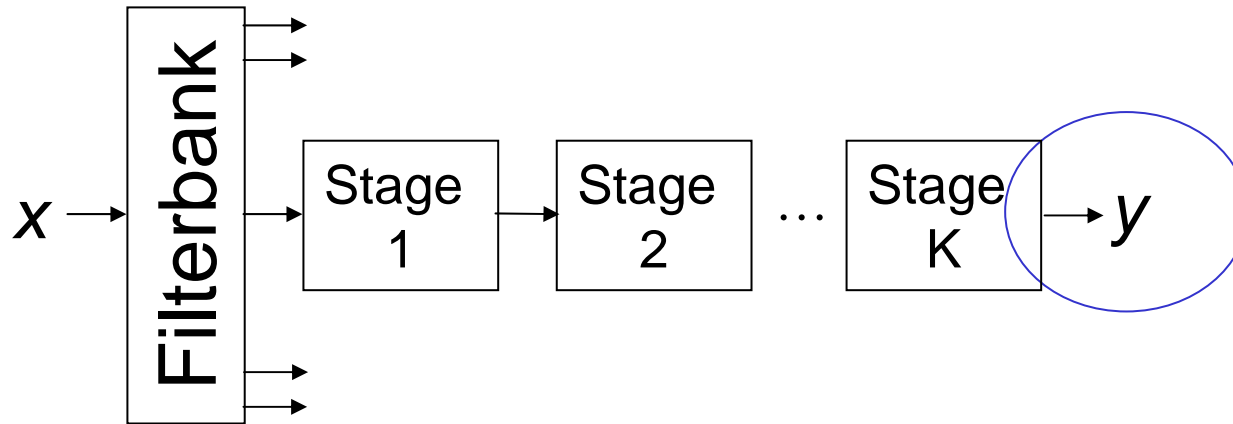
➤ **Analysis** of model properties using **linear algebra**

- Masking curve
- Subspace-based analysis

- A typical auditory model (here: the Dau model)



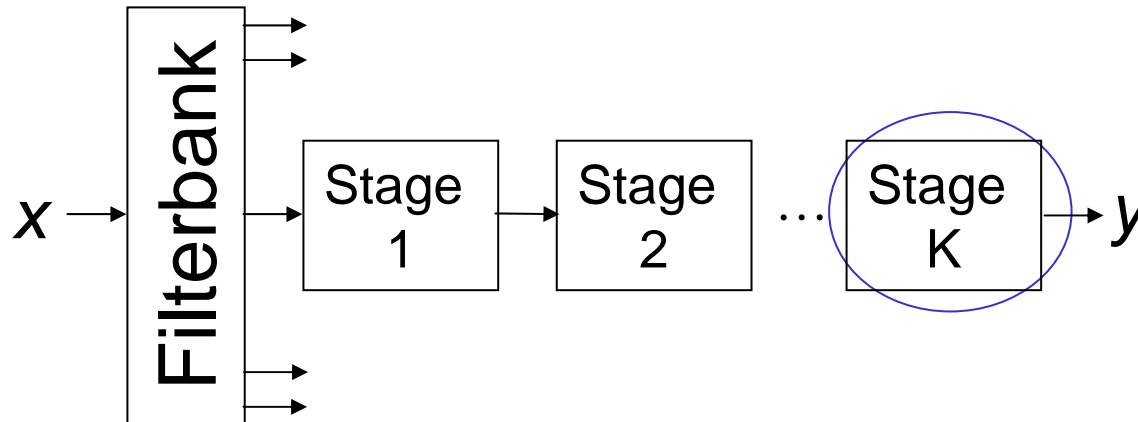
- Distortion is sum of per-channel distortions



- Find **sensitivity matrix** for model output y per channel m

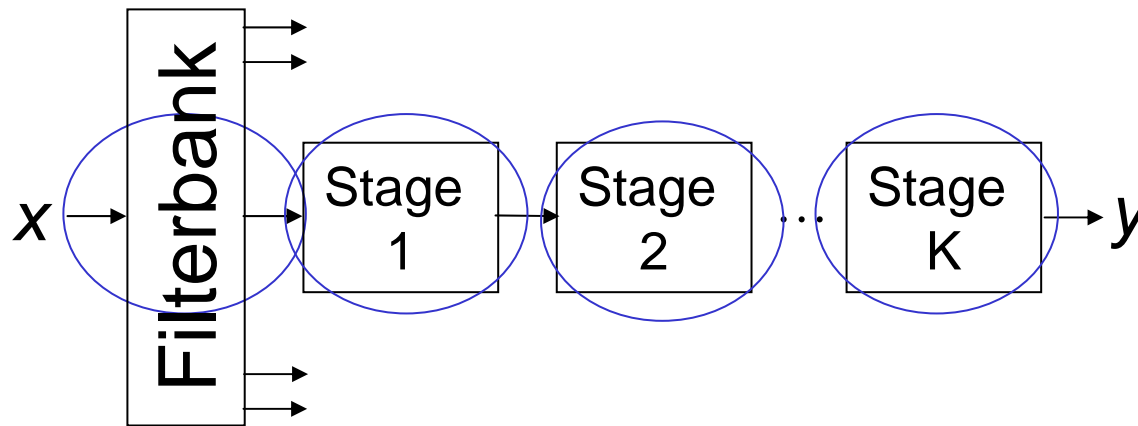
$$d^{(m)}(\mathbf{y}, \hat{\mathbf{y}}) = \frac{1}{2} (\mathbf{y} - \hat{\mathbf{y}})^H \mathbf{M}_y^{(m)}(\mathbf{y}) (\mathbf{y} - \hat{\mathbf{y}})$$

here $\mathbf{M}_y^{(m)}(\mathbf{y}) = 2\mathbf{I}$



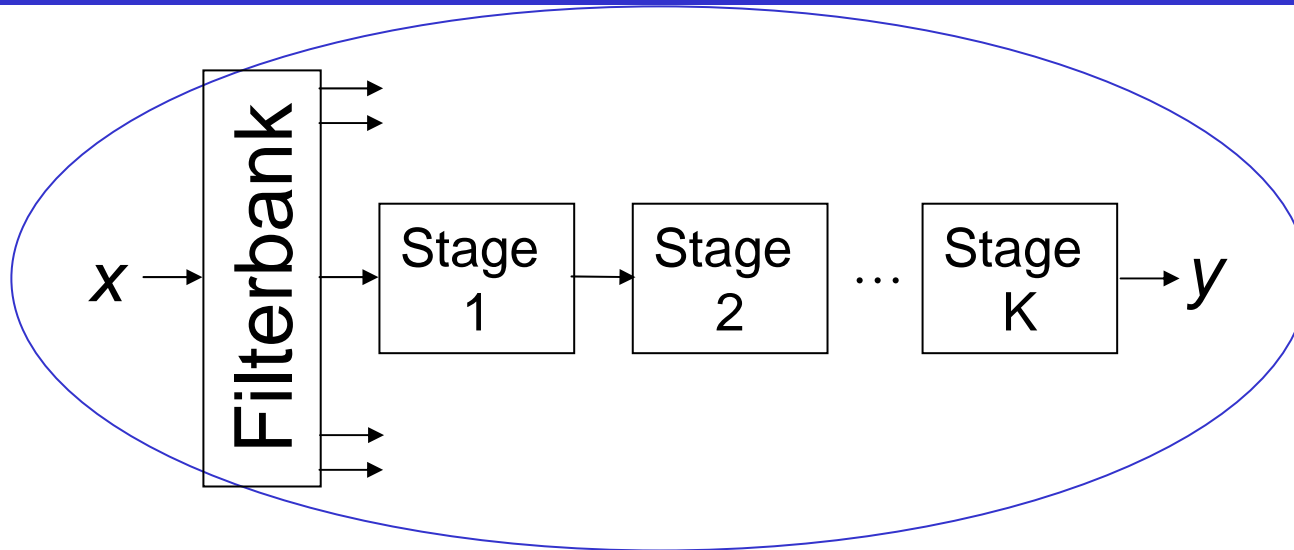
- Each model **stage** linearized by respective Jacobian

$$\partial \mathbf{z} \longrightarrow \mathbf{J}_z(\mathbf{z}) \longrightarrow \partial \mathbf{y}$$



- Get sensitivity matrix for channel m in input signal domain from **chain rule**;

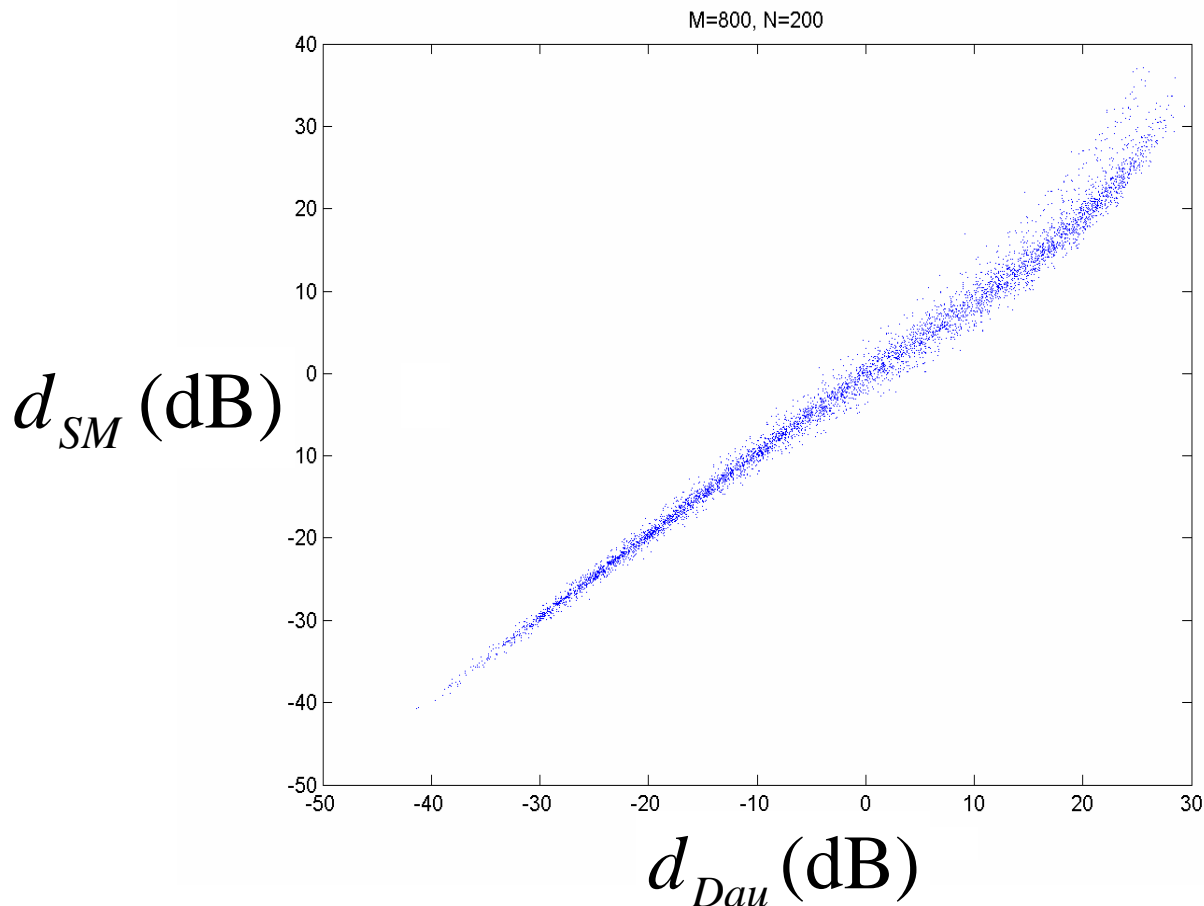
$$\mathbf{M}_x^{(m)}(x) = 2 \left[\prod_k \mathbf{J}_k^{(m)} \right]^H \prod_k \mathbf{J}_k^{(m)}$$



- **Sensitivity Matrix** is sum of per-channel matrices

$$\mathbf{M}_x(\mathbf{x}) = \sum_m \mathbf{M}_x^{(m)}(\mathbf{x}) = 2 \sum_m \left[\prod_k \mathbf{J}_k^{(m)} \right]^H \prod_k \mathbf{J}_k^{(m)}$$

- Narrow-band music + white noise at different SNR (10-60 dB), correlation > 0.9



coding
possible


- Sensitivity matrix $\mathbf{M}_X(\mathbf{X})$ for frequency domain vectors, with \mathbf{F} = DFT in matrix notation

$$(\mathbf{x} - \hat{\mathbf{x}}) = \mathbf{F}^H (\mathbf{X} - \hat{\mathbf{X}})$$

$$\mathbf{M}_X(\mathbf{X}) = \mathbf{F} \mathbf{M}_x(\mathbf{x}) \mathbf{F}^H$$

- Assume **masking threshold** is at $d(\mathbf{X}, \hat{\mathbf{X}}) = 1$
- **Masking curve** at frequency i :
gain g_i for unit distortion in DFT bin i

$$\mathbf{u}_i = [0 \quad \dots \quad 1 \quad \dots \quad 0]^H$$

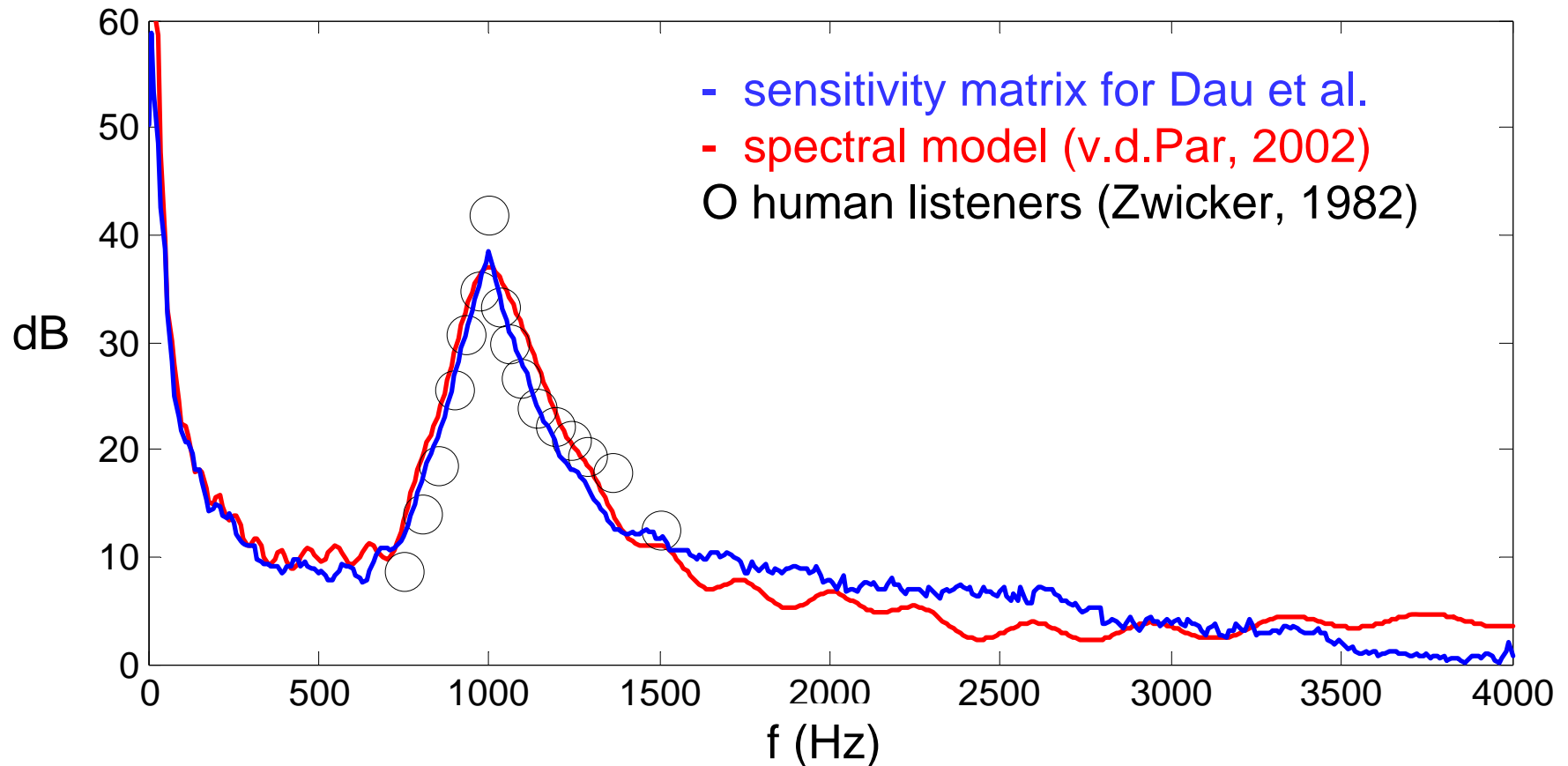

 pos. i

to approach masking threshold

$$1 = \frac{1}{2} \mathbf{u}_i^H \mathbf{M}_X(\mathbf{X}) \mathbf{u}_i g_i$$

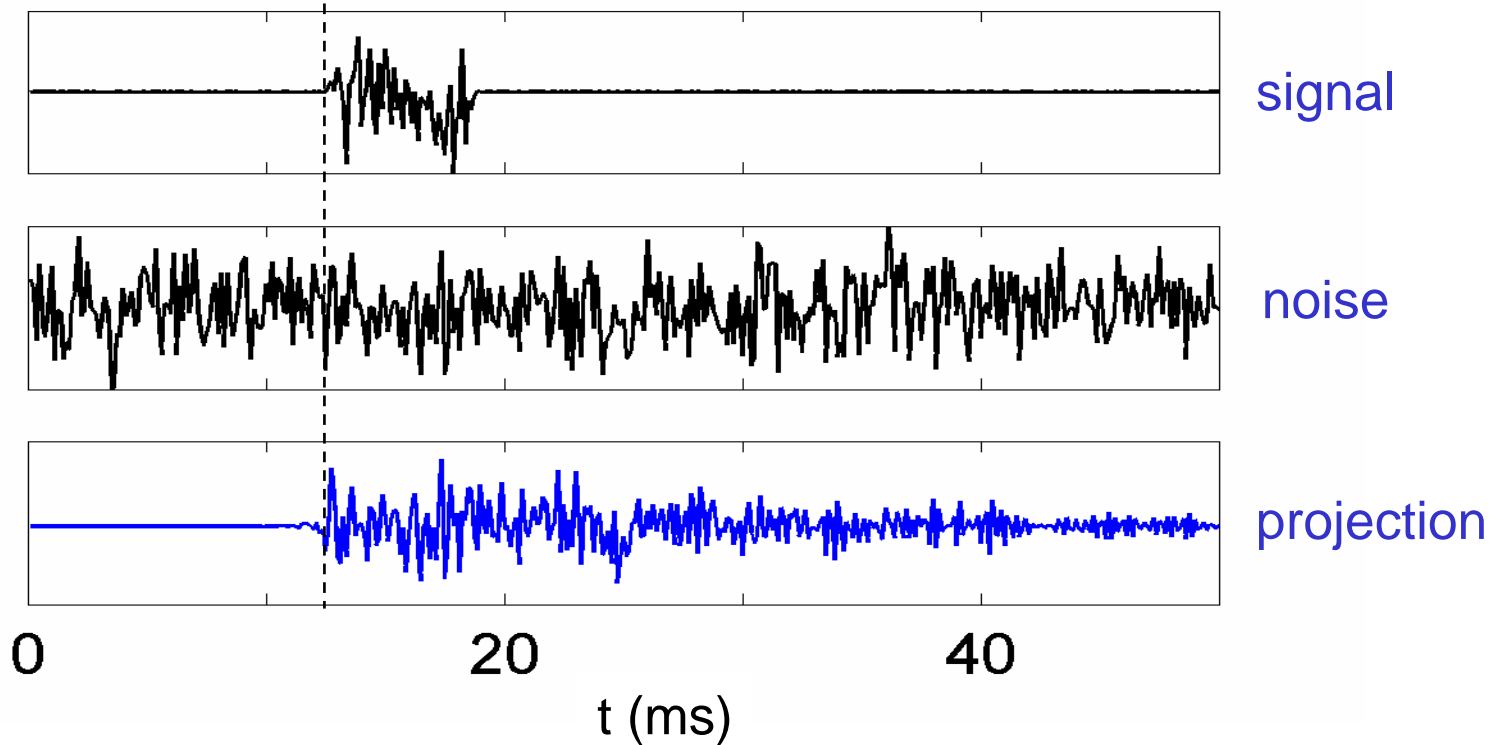
$$\Leftrightarrow g_i = \frac{2}{[\mathbf{M}_X(\mathbf{X})]_{i,i}}$$

- Masking curve for 50 dB sinusoid at 1000 Hz

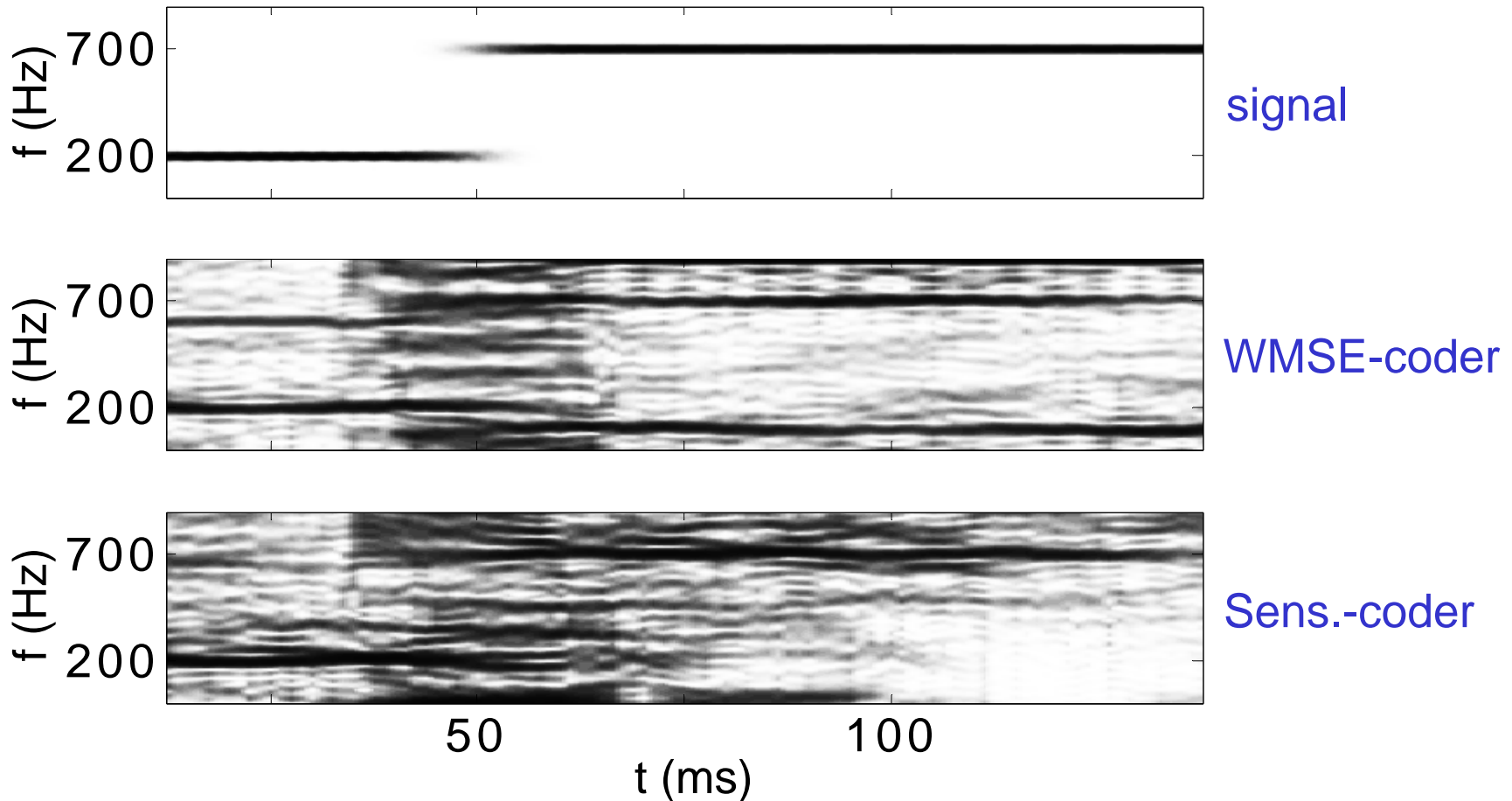


FlexCode Example 2: Non-Simultaneous Masking

- Subspace analysis of **M**: low sensitivity space



- Spectrograms for two sinusoids (200 Hz and 700 Hz) coded with an APCM coder



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- Sensitivity Matrix generally **not diagonal** in typical coding domains (e.g., time, frequency, etc.)
 - distortion **not single-letter**:

$$d(\mathbf{x}, \hat{\mathbf{x}}) \approx \frac{1}{2} (\mathbf{x} - \hat{\mathbf{x}})^H \mathbf{M}_x(\mathbf{x}) (\mathbf{x} - \hat{\mathbf{x}})$$

- no **analytical** solution known for **optimal** coding
- so far only **trained VQ** available (undoable for large dimensions)!

- Sensitivity Matrix **diagonal** in perceptual domain
 - **transform** signal vectors into perceptual domain
 - **(W)MSE-optimal coding** in perceptual domain results in minimum perceptual distortion
- Optimal Transform:

$$\mathbf{F}'_{opt}{}^T(\mathbf{x})\mathbf{F}'_{opt}(\mathbf{x}) = c\mathbf{M}_x(\mathbf{x})$$

- **Problem:** optimal transform depends on \mathbf{x}
 - not known at decoder!
- **Solution:** Parameterize transform as DCT + diagonal weights

$$\mathbf{F}'(\mathbf{x}) = \mathbf{W}_f \mathbf{T}_{DCT} \mathbf{W}_t$$

- combines DCT coding gain with perceptual transform!
- Model-driven approach to temporal noise shaping (TNS)

- Original 
- Freq. weights  $\mathbf{F}'(\mathbf{x}) = \mathbf{W}_f \mathbf{T}_{DCT}$
- Time-freq. weights  $\mathbf{F}'(\mathbf{x}) = \mathbf{W}_f \mathbf{T}_{DCT} \mathbf{W}_t$

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- Main complexity lies here:

$$\mathbf{M}(\mathbf{x}) = 2 \sum_m \left[\prod_k \mathbf{J}_k^{(m)} \right]^H \prod_k \mathbf{J}_k^{(m)}$$

- Can we make multiplications ‘sparse’?
- Can we reduce the size of the matrices?

- Using auditory models in coding is possible
- General method
- Direct applicability in AbS structures (CELP)
- Solution outline for transform coding

Thank you!