

# Master MVA

## Analyse des signaux Audiofréquences

*Audio Signal Analysis, Indexing and Transformation*

### Lecture on Sound effects and Reverberation

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# Content

## ➤ Sound effects

- Compression/Expansion – Limitation
- Flanging – Phasing – Wah/wah
- Time-frequency modifications
  - Definition
  - Circular memory technic
  - PSOLA
- Distortion

## ➤ Artificial reverberation

- Introduction
- Algorithms (perceptual, physics based)



# Compression/Expansion – Limitation

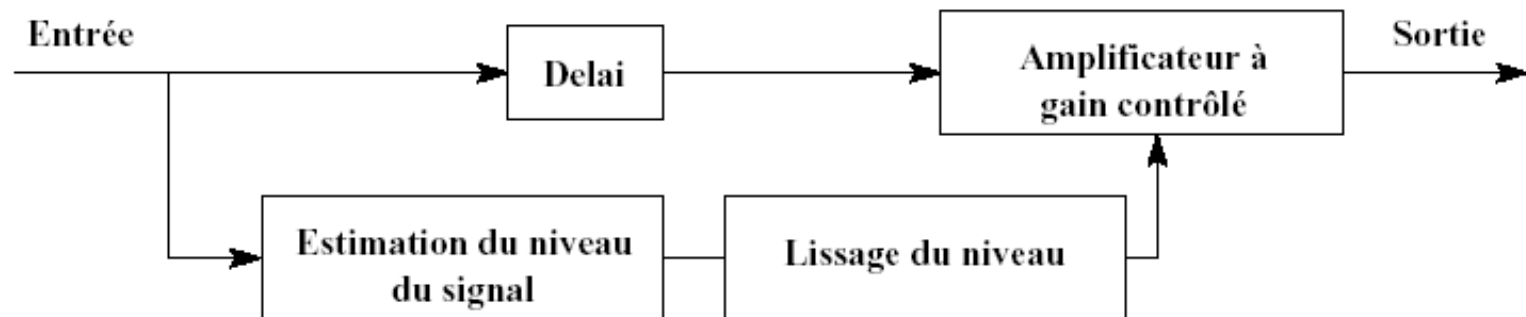
- Largely used in analog systems to control the dynamic range of signals
  - **Dynamic range of signals:** ratio between the power of the louder signal to the power of the weakest signal
- **Order of magnitude:**
  - Dynamic of a room/hall:  $> 100$  dB
  - Dynamic of a FM radio broadcast :  $< 70$  dB
- **Principle:**
  - Compression before transmission
  - Expansion after transmission to restore the initial dynamic



# Compression/Expansion – Limitation

## ■ Principe:

- Estimate the level of the input signal
- Adjust the parameters of the dynamic control
- Signal level modification according to the dynamic control gain (in logarithm scale).



# Estimation of the input signal level

- Root-mean square (RMS) value:

$$P_{eff} = \sqrt{\frac{1}{T} \int_{t-T}^t x^2(u) du}$$

$$P_{eff} = \sqrt{\frac{1}{N} \sum_{i=n-N+1}^n x_i^2}$$

- Peak to Peak level :

$$P_{crete} = \max_{n-M < i < n} |x_i|$$

- For a sinusoidal signal:

$$P_{eff} = A/\sqrt{2}$$

$$P_{crete} = A$$



# Estimation of the input signal level

## ■ Crest factor: ratio of the peak to peak level and the root mean square value

- Always greater than 1
- Measures the presence of peaks (e.g. impulses) in the signal
- For a sinusoid: Crest factor =  $P_{crete} / P_{eff} = \sqrt{2}$

## ■ Estimation of input signal level

- In RMS for **compression/expansion**,
- Peak-to-peak value for **limitation**.



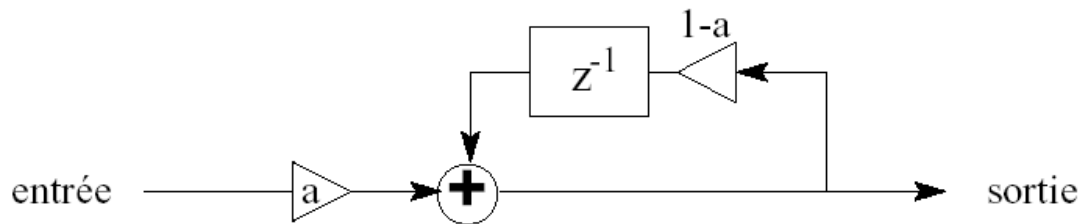
# Real-time update of dynamic control parameters

- **Input level smoothing** to avoid too abrupt variations of the amplification gain.
- **Dynamic control: 2 parameters:**
  - **Rising time (or Attack time):** How soon the compressor starts to compress the dynamics after the threshold is exceeded.
    - *If volume changes are slow, a high value is possible. Short attack times will result in a fast response to sudden, loud sounds, but will make the changes in volume much more obvious to listeners.*
  - **Release time:** How soon the compressor starts to release the volume level back to normal after the level drops below the threshold.
    - *A long time value will tend to loose quiet sounds that come after loud ones, but will avoid the volume being raised too much during short quiet sections like pauses in speech.*



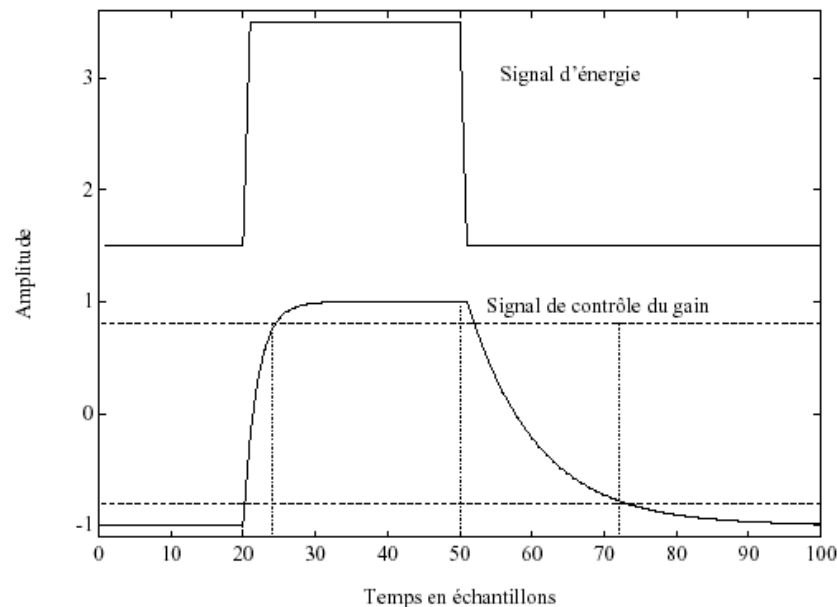
# Real-time update of dynamic control parameters

## ■ Low-pass implementation



$$H(z) = \frac{a}{1 - (1 - a)z^{-1}}$$

## ■ Examples:

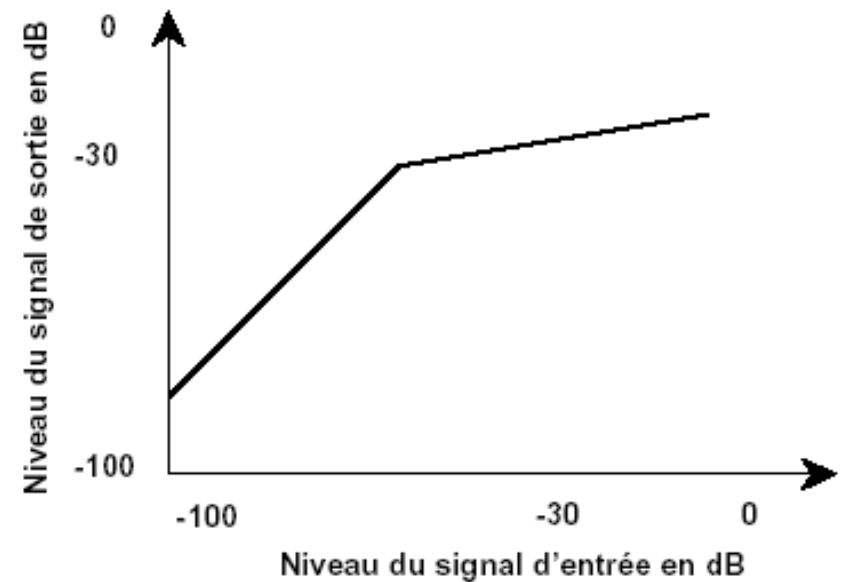
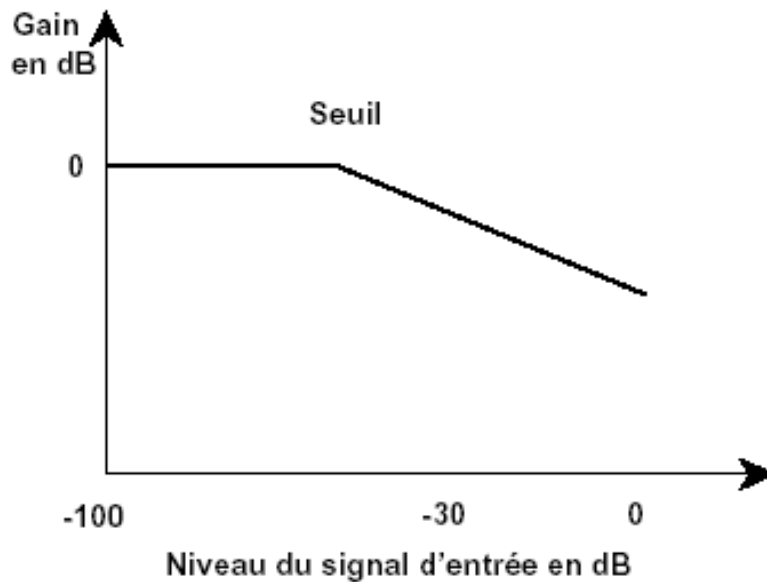




# Dynamic compression (or Compressor)

## ■ Compression is characterized by :

- A threshold
- A compression ratio



# Dynamic compression (or Compressor)

- Amplification gain is given by:

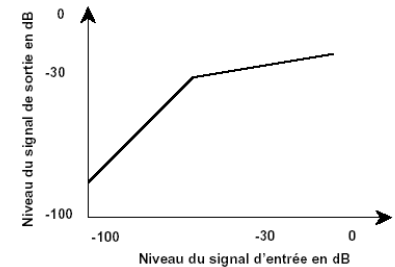
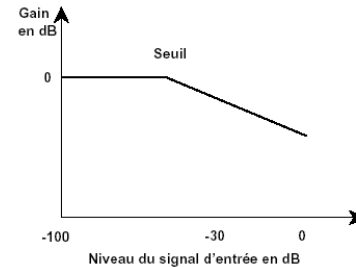
$$Y_{eff} = X_{eff} + g(X_{eff})$$

Where  $\alpha$  is the gain slope

$$\frac{dY_{eff}}{dX_{eff}} = 1 + \frac{dg(X_{eff})}{dX_{eff}} = 1 + \alpha$$

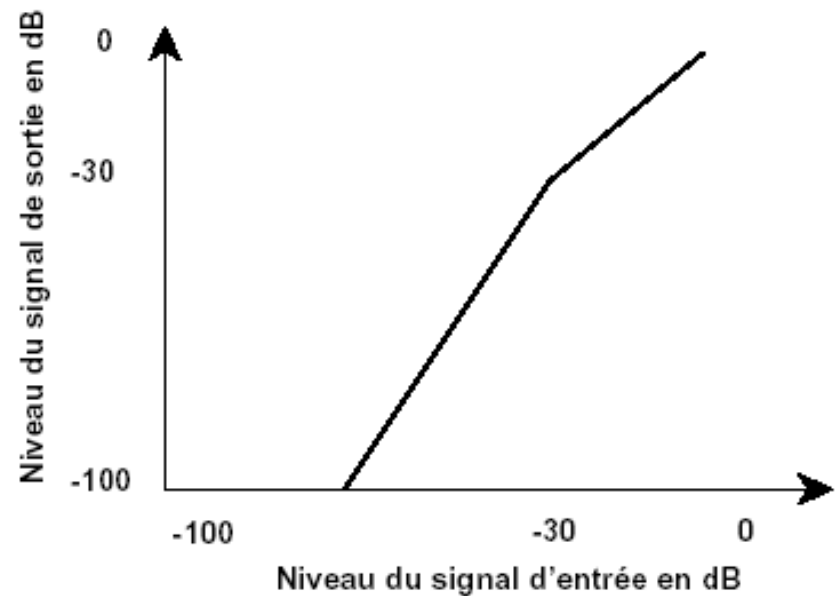
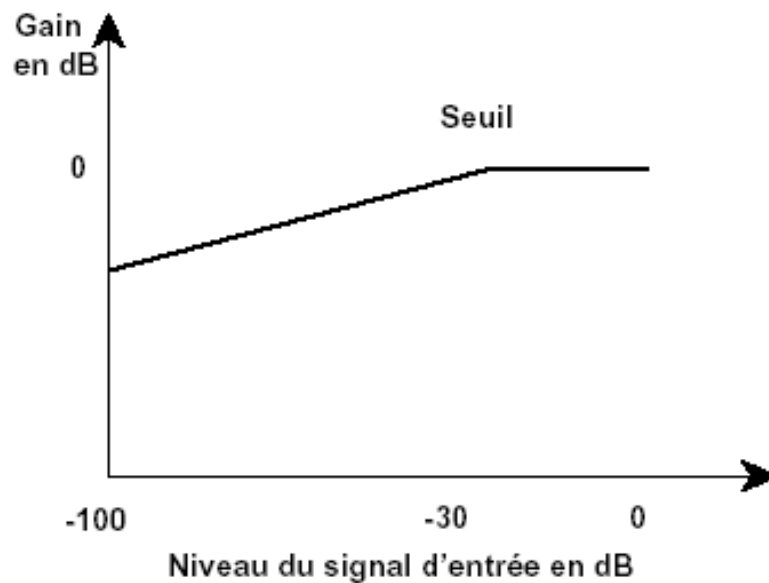
If the gain is linear in  $X_{eff}$  with slope  $\alpha$ , the relation input/output is also linear with a slope equal to  $1 + \alpha$

In practice : ratio of from 2:1 to 10:1



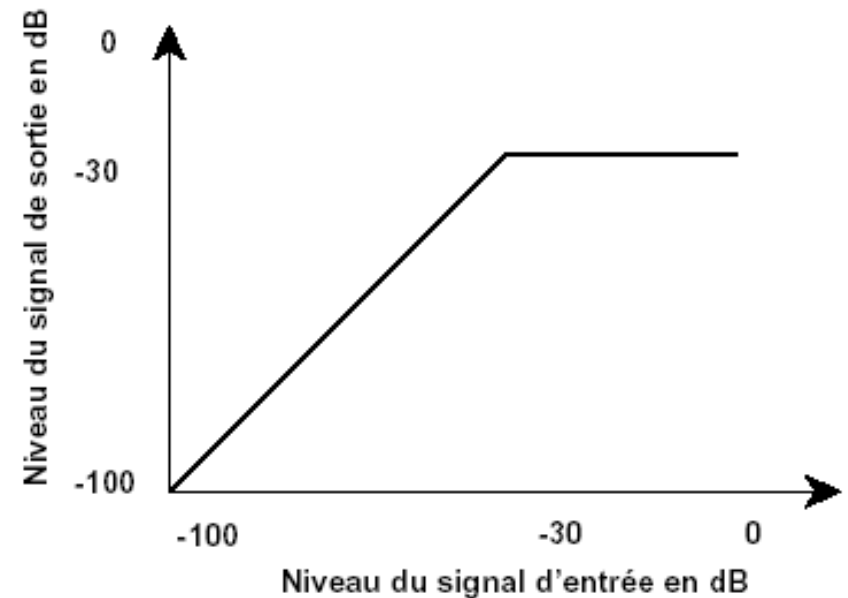
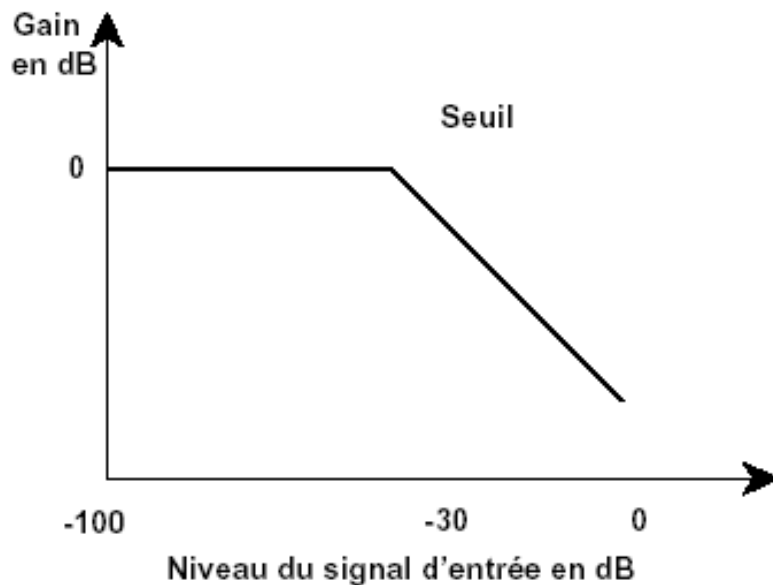
# Dynamic expansion (Expander)

## ■ Expansion characteristics



# Limitation (or Limiter)

## ■ Limitation characteristics



# Use cases

- Background noise 'suppression':
  - Dolby B, C...are based on reversible compression/expansion
- Recording of a signal with high dynamic (CD) on a traditional lower dynamic support (tape)
- High gain compression for producing a sustain effect (well appreciated by guitarists...)
- High compression used by some radio .. which broadcast at high mean average level....
- Limitation used to avoid clipping, distortion (which are more detrimental to the signal quality than the effect of limitation).



# Flanging / Phasing

## ■ Origin:

- Use of the thumb on two turntables playing the same tune.
- **The thumb effect** : slows down one piece compared to the other one which are then sometimes synchronised and sometimes slightly out of phase.

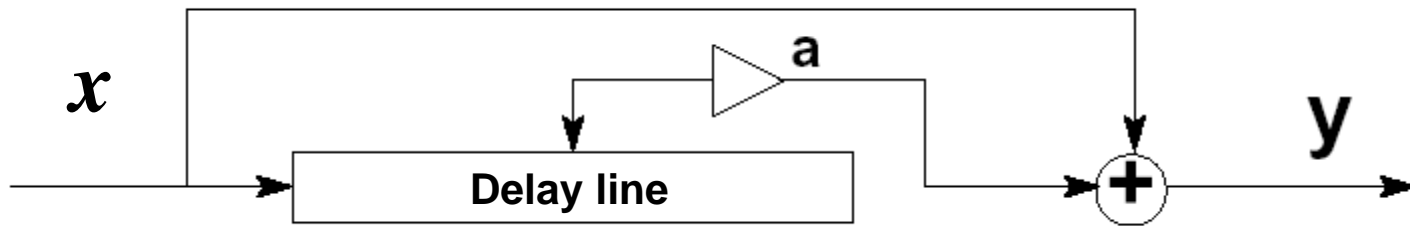
## ■ Interpretation:

- The signals coming from the two turntables are delayed (with a variable delay)
- When the two signals are added some sinusoidal components are cancelled because they are in opposition of phase (e.g. *phase inverted*)
- **Effect:** Flanging then introduces “holes” in the spectrum (regularly spaced) and their positions change in time.



# Flanging / Phasing

## ■ Implementation



## □ Transfer function

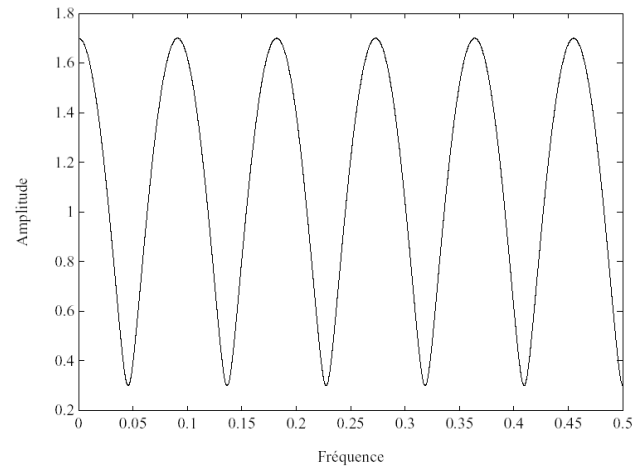
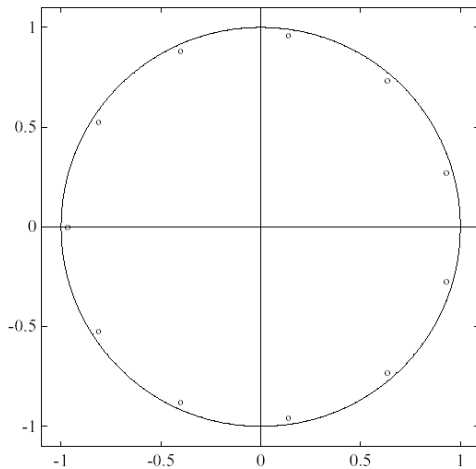
$$H(z) = 1 + a z^{-p} \qquad |H(e^{j\omega})| = \sqrt{1 + a^2 + 2a \cos(p\omega)}$$

## □ $p$ minima at odd multiples of $F_e/2p$



# Flanging / Phasing

## ■ Transfer function of the « fixed » (non variable) system



- This transfer function has a meaning only for slowly variations of the delay  $p$  ....
- Strength of the effect: parameter  $a$

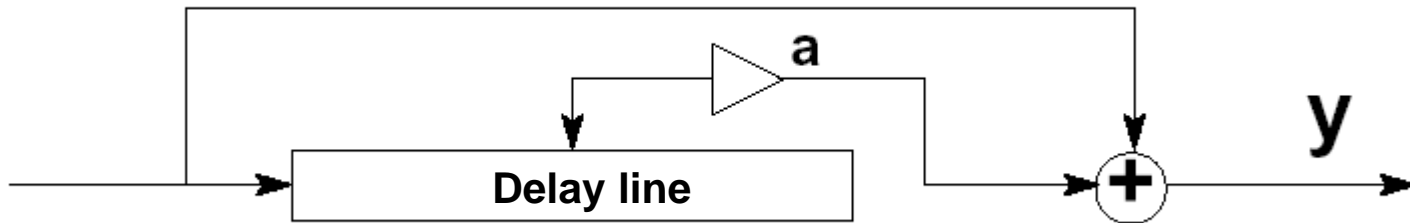
$$R_{dB} = 20 (\log_{10}(1 + a) - \log_{10}(1 - a))$$





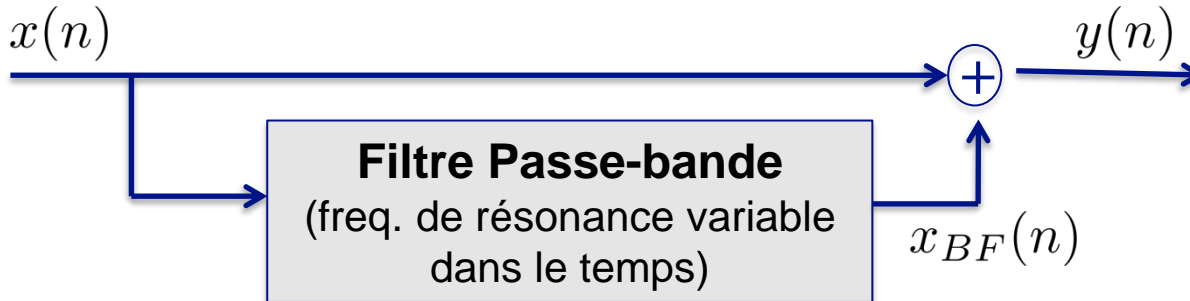
# Phasing

- Phasing is based on a variable comb filter but where here the gain  $a$  is variable (not the delay)



# “Wha-wha” effect

- Effect “Wha-wha”: The delay (e.g. notch filter) is replaced by a time varying bandpass filter :



- **Demonstration:**

*Original*



*Wah-wah (modulation of resonance frequency with a triangle function between 500Hz and 3000Hz)*



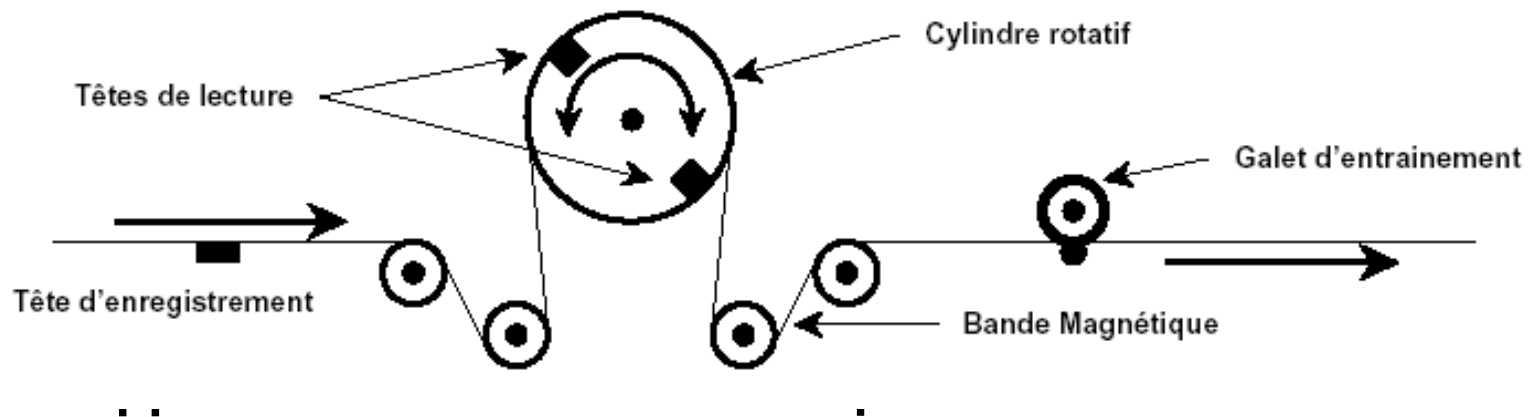
# Time and pitch shifting

- **Time-shifting** : to modify the duration of an audio recording without affecting the frequency content.
- **Pitch shifting**: to modify the frequency content (e.g pitch) of an audio recording without affecting its duration.
  - Pitch shifting without modifying the main resonances (e.g. formant for speech).
  - Straight pitch shifting (e.g. auto tune etc ..)
- **Various methods exist**
  - TD-Psola, Phase vocoders
  - Specific audio methods



# Circular memory technique

## ■ The old times ...

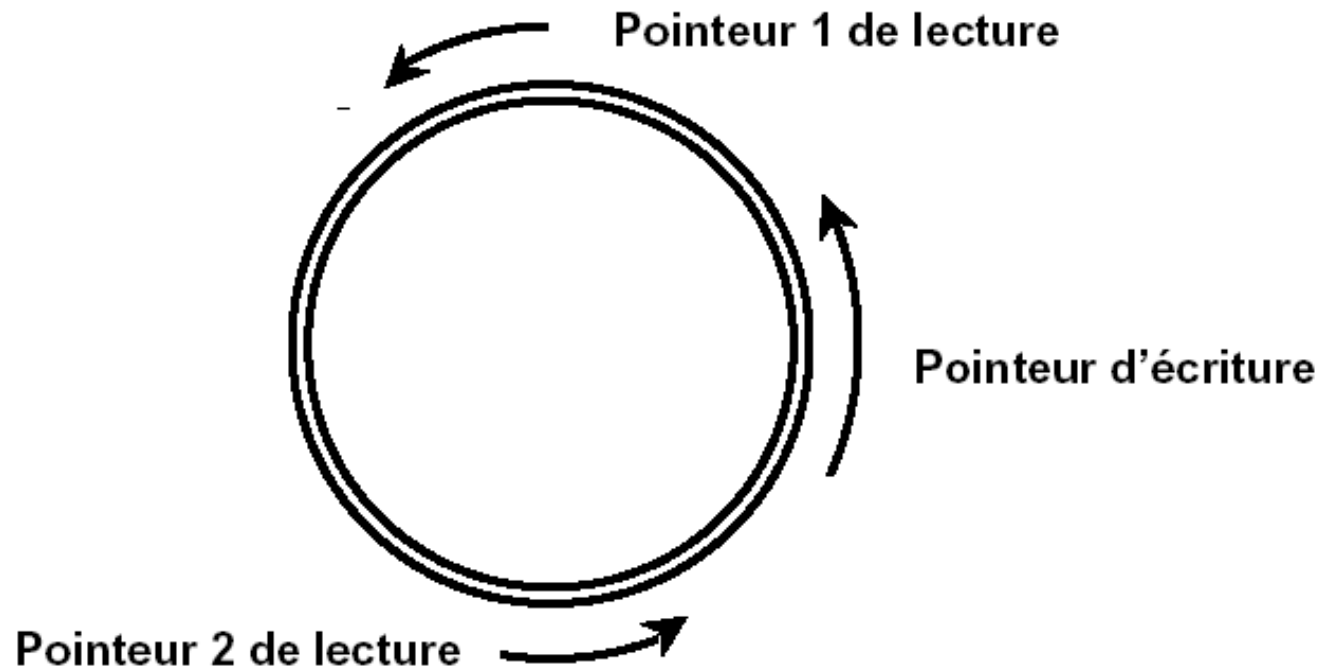


$$\alpha = \frac{V_r}{V_a} = \frac{V_a + R \Omega_{cylindre}}{V_a}$$



# Circular memory technique

## ■ In digital ...





# Circular memory technique

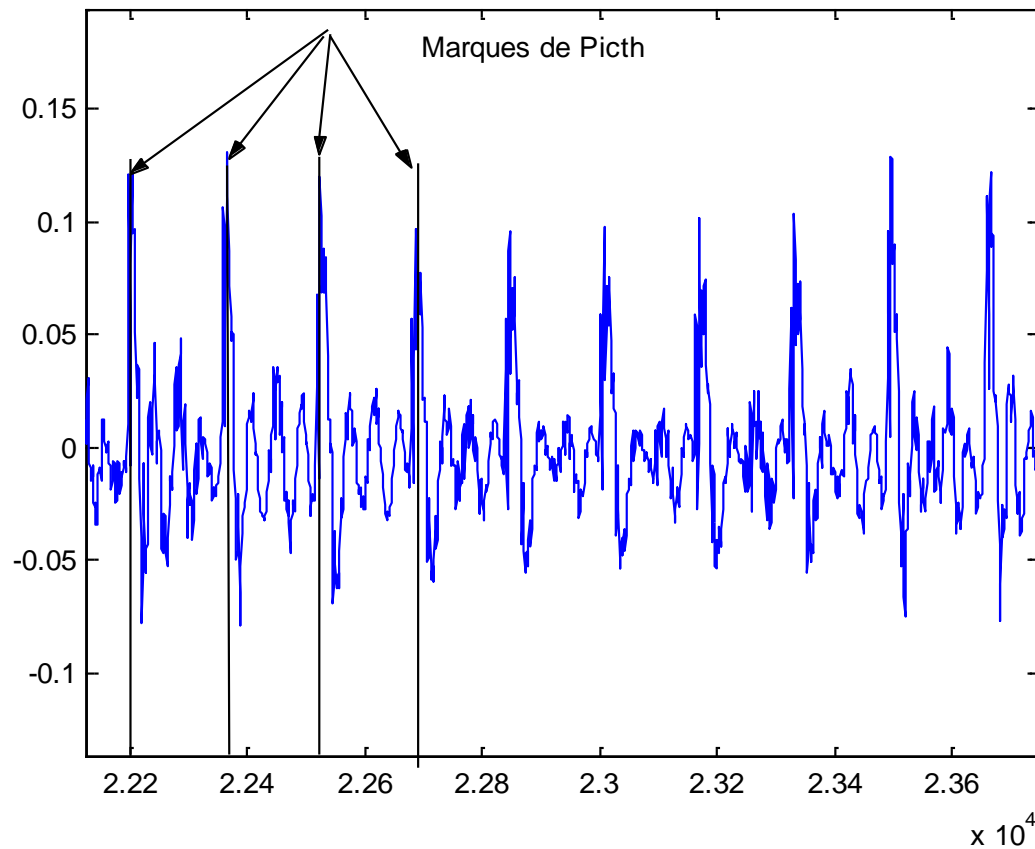
## ■ Time-shifting

- Change the sampling frequency by a factor  $\alpha$
- Change the pitch using the circular memory technique
- The two operations can be done simultaneously (we now get closer to the TD-PSOLA approach)

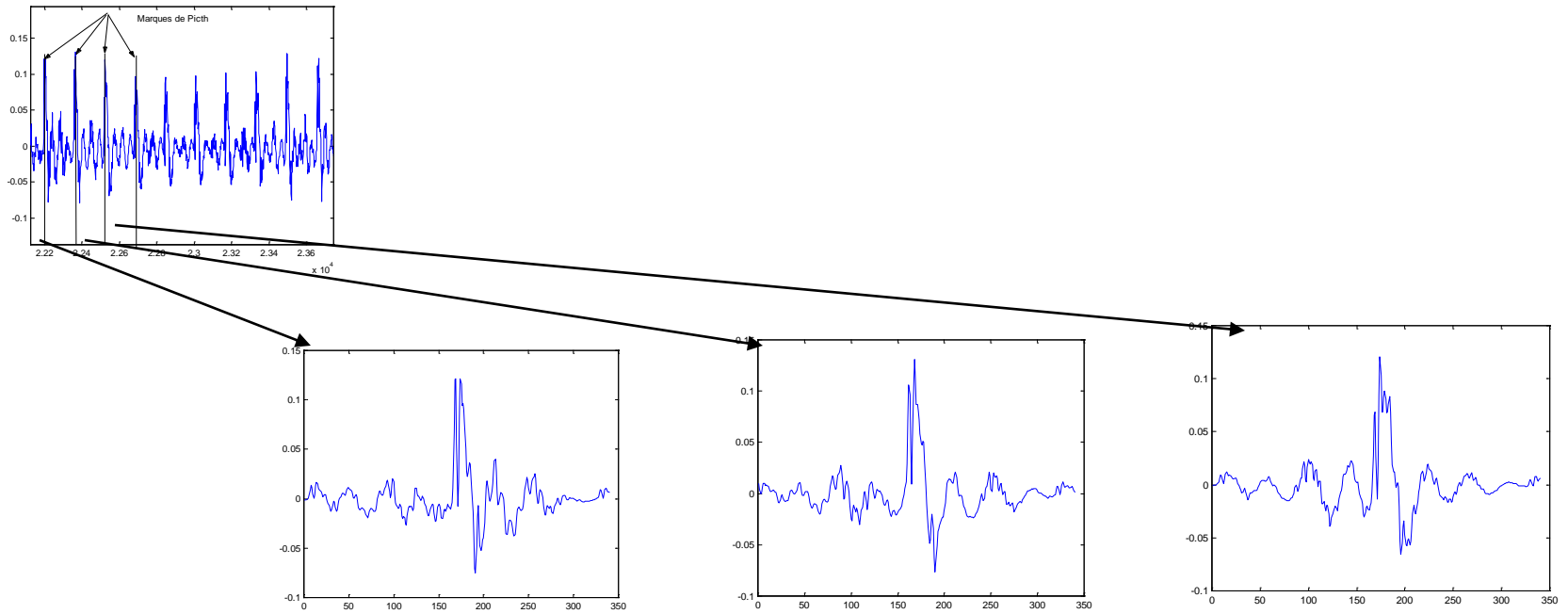


# TD-PSOLA

## TD-PSOLA: Time Domain Pitch Synchronous OverLap and Add

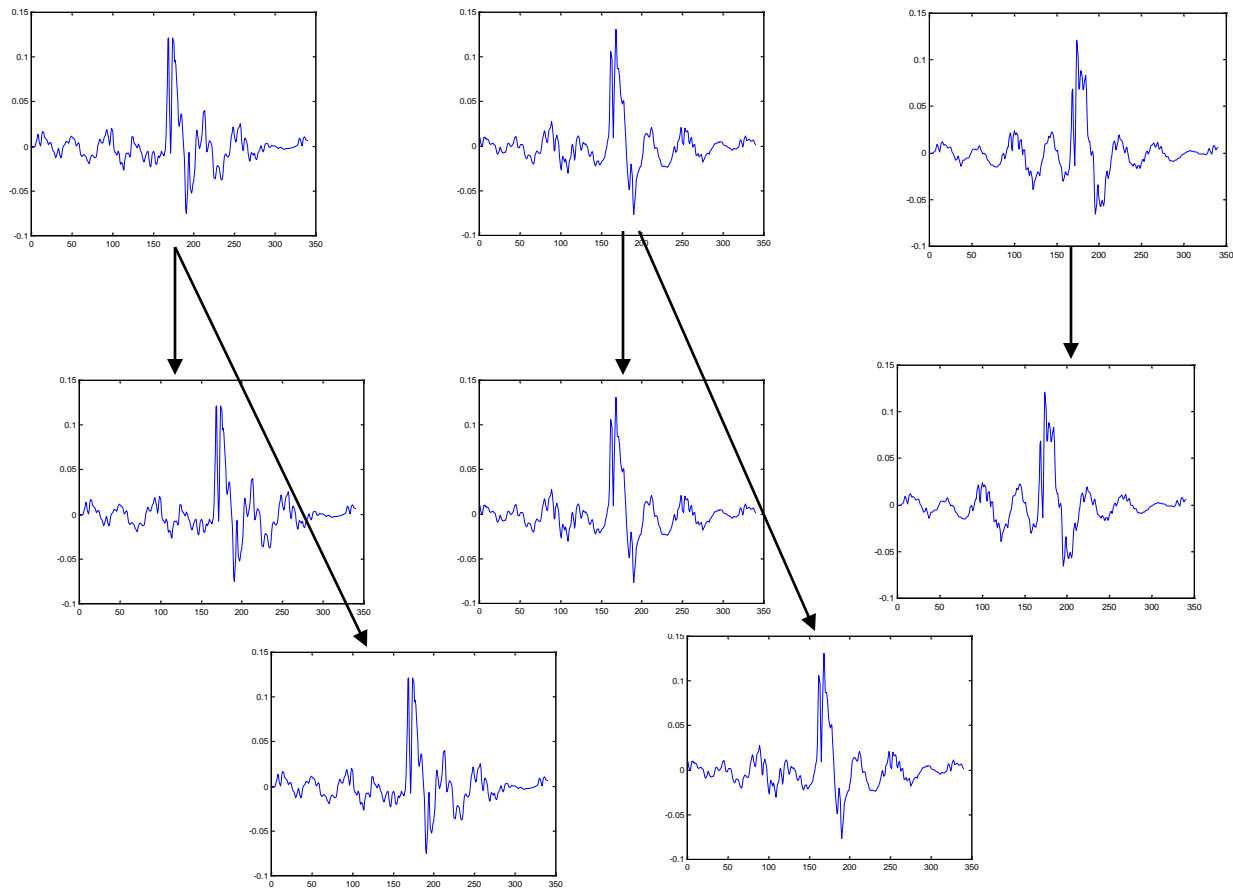


# Short-term signals

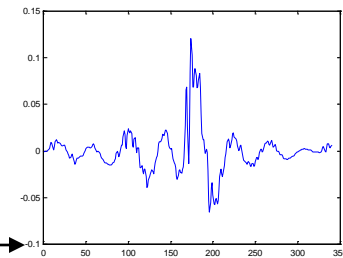
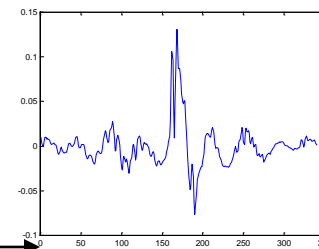
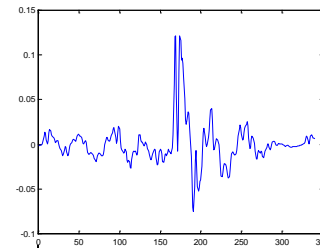
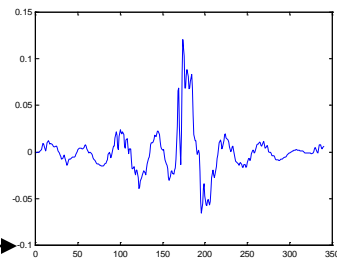
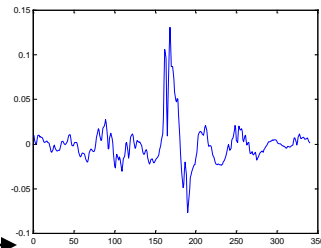
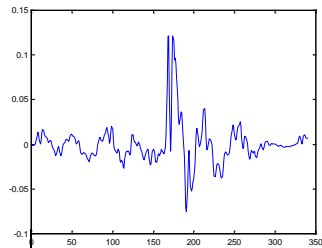




# Time-shifting



# Pitch shifting



# TD-PSOLA: Synthesis

## ■ Extraction of pitch synchronous short term waveforms

- Analysis window spanning 2 periods of the analysed signal (for quasi-periodic segments).
- Fixed duration analysis window (around 10ms) for more complex segment (noise or high number of notes).

## ■ Synthesis by Overlap and Add

- Insertion / Suppression of short term waveforms to modify duration
- Modification of short term waveforms spacing to modify pitch.

## ■ Sound example (speech) –



Original



accéléré (psola)



lecture accélérée

## ■ Sound example (singing voice)



*Transposed*





# Distorsion

- Usually, distorsion is not desired ....
- But it could also be an important artistic effect (for guitarist, for instance....)
- Distortion is traditionally obtained by applying a non-linear function to the audio signal.



# Distorsion

- It can be shown that distortion gives birth to new components in the signal :

- The development of the non-linear function is given by:

$$f(u) = f(0) + uf'(0) + \frac{u^2 f''(0)}{2!} + \frac{u^3 f'''(0)}{3!} + \dots = \sum_{i=0}^{\infty} \frac{u^i}{i!} \frac{d^i f}{du^i}$$

- Applying the fonction  $f$  to the audio signal  $x_n = \sum_{i=-p}^p e^{j\omega_i n}$

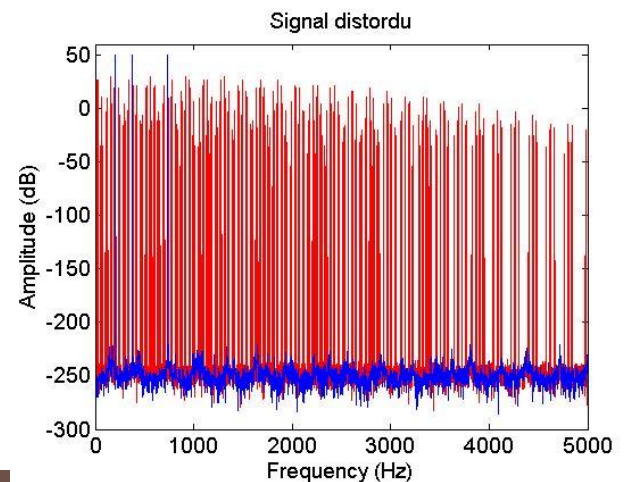
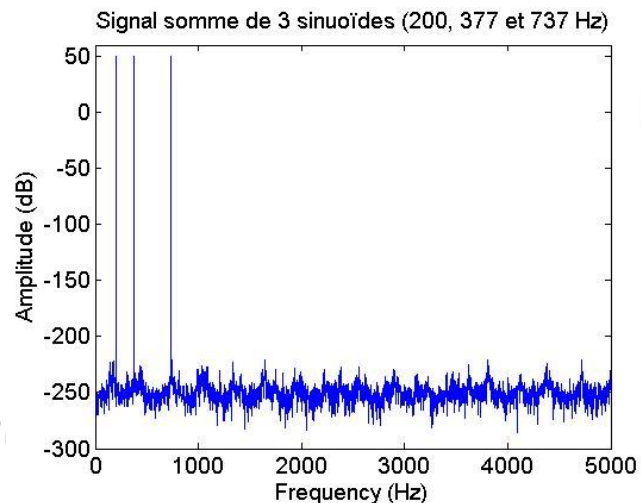
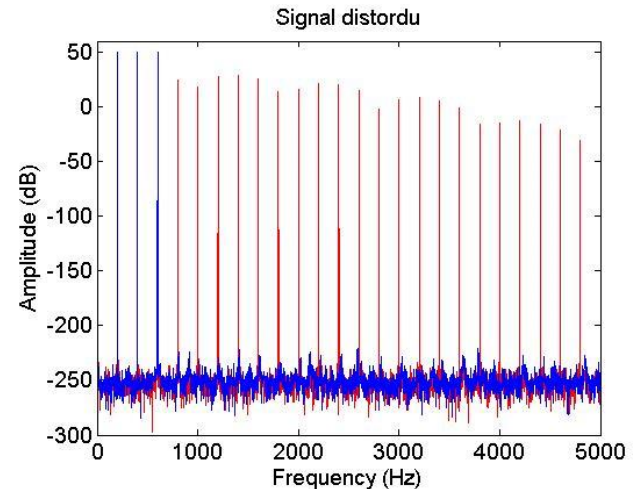
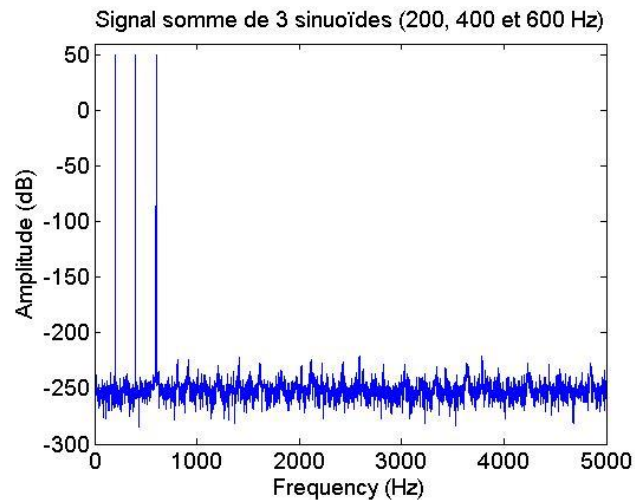
results in a weighted sum of its integer  $y_n = \sum_{i=0}^{\infty} \alpha_i (x_n)^i$

with  $(x_n)^i = \sum_{-p \leq j(k) \leq p} e^{j(\omega_{j(0)} + \omega_{j(1)} + \dots + \omega_{j(i-1)})}$



# Examples

- Non linear function:  $f(x) = x^9$  with  $x = \sum_{i=1}^3 A_i \sin(2\pi f_i t)$



# Sound effects with deep learning

*A growing trend*

## ■ Two main approaches

- As parameter estimators of audio processors
  - DNN to predict dynamic range processors parameters [Ram19]
  - Filter gain estimation for EQ with multilayer perceptron [She19]
- As end-to-end methods
  - Virtual analog models of nonlinear effects with DNN [Ram20a]
  - Deep learning for Audio effects modeling [Ram20b]
  - Learning differential equations with DNN [Wil22]

[Ram19] J. Rämö and V. Välimäki. Neural third-octave graphic equalizer. In Proc. of DAFx-19, 2019.

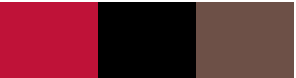
[She19] D. Sheng and G. Fazekas. A feature learning siamese model for intelligent control of the dynamic range compressor. In Proc. of IJCNN, 2019 .

[Ram20a] M. Ramírez, E. Benetos, J. Reiss Deep Learning for Black-Box Modeling of Audio Effects. *Applied Sciences*. 2020

[Ram20b] M. Ramírez, Deep learning for Audio effects modeling, PhD thesis, Queen Mary Univ., UK 2020.

[Wil22] J. Wilczek & al. Virtual Analog modeling of distortion circuits using neural ordinary differential equations , in Proc. of Dafx22, Vienna, 2022.





# Artificial reverberation



Droits d'usage autorisé

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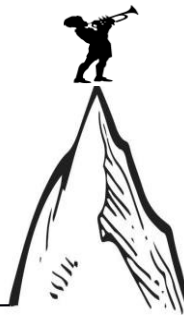


## Situations with no reverberation

- When in an anechoic room ...



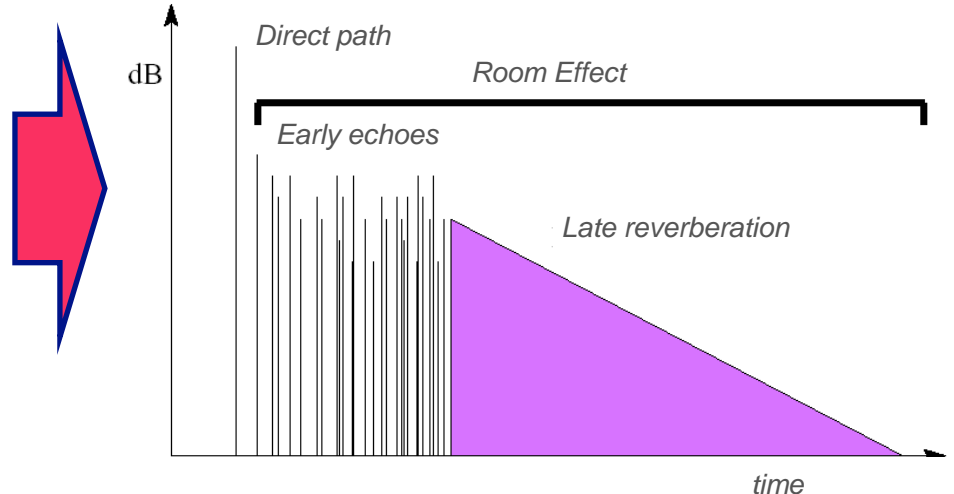
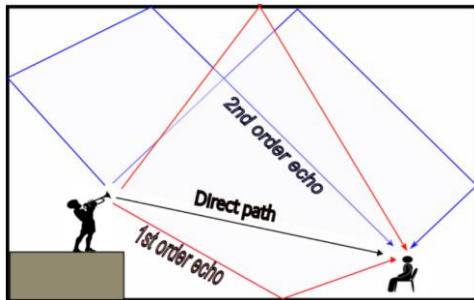
- .. Or when in “free field”



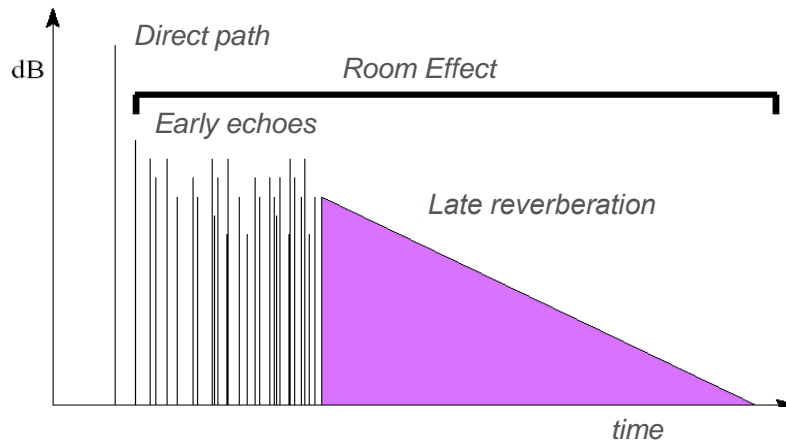
# Reverberation: Room effect

## ■ Room effect can be decomposed in:

- A contribution due to **early echoes** or early reflexions (which depends on the room geometry and on the positions of the source and microphone)
- A contribution due to **late reverberation** (which mainly depends on the volume and global absorption of the room)



# Reverberation: Room effect



## ■ Room effect = filtering effect

$$y(t) = \int_0^{\infty} x(t - u)h(u)du$$

• or

$$y(n) = \sum_{i=0}^{\infty} x(n - i)h(i)$$

The Room Impulse Response (RIR)  
(or acoustic channel)



# Applications: Reverberation and Dereverberation

- **Artificial reverberation : generating a new signal with different reverberation characteristics:**

$$\hat{y}(n) = \sum_{i=0}^{\infty} x(n-i)h(i)$$

— Applications:

- Studio recordings and mixing
- Live music (reverb pedals, synthesizers,...)
- Virtual reality and movie production



- **Dereverberation: removing the reverberation effect to retrieve the original source (or « dry » signal)**

”Recovering  $\hat{x}(n)$  from the reverberated signal  $y(n)$ ”

— Applications:

- Speech enhancement (especially late reverberation removal to increase intelligibility)
- Robust speech recognition
- Acoustic transfer



# Artificial reverberation: a long history ...

(from [1, 2])

## ■ From analog devices in 1920's ...

*(..transmit the sound into an empty acoustic space, and recording the response of the space via a microphone..)*

- .. to spring resonator (late 1920's)..
- ..to plate reverberator, such as the EMT140 (in the 1950's)
- .. to "Bucket-Brigade" Device (BBD), by Philips (in the 1960's)

## ■ .. to Digital methods ..

- ... delay networks (as Schroeder reverberator in late 1960's) : (..the input signal is delayed, filtered and fed back along a number of paths according to parametrized reverberation characteristics)
- ... convolutional (typically, the input signal is convolved with a recorded or estimated impulse response of an acoustic space)
- ... physical models (typically the input signal drives a simulation of acoustic energy propagation in the modeled geometry).

## ■ ... to deep learning (e.g. data based) methods

- ..for instance learning the parameters of a reverberation model using deep learning (as in [3])



# Artificial reverberation

## *the interest for Hybrid methods*

### ■ Physics-based methods

- Accurate sound propagation modeling
- Relatively high complexity
  - Image source method
  - Radiance transfer method
  - Beam tracing method, . . .

### ■ « Perception »-based methods

- Computational efficiency
- Not based on room geometry
  - Schroeder reverberation model
  - Feedback delay networks, . . .

### ■ The interest for Hybrid methods

- Can link geometry-based models and perception-based models [1]
- Can exploit machine learning (deep learning) to learn model parameters [3]



[1] H. Bai, G. Richard, and L. Daudet. "Geometric-based reverberator using acoustic rendering networks." In Proceedings of the IEEE Workshop on Applications of Signal Processing to Audio and Acoustics (WASPAA), pages 1–4, New Paltz, NY, October 2015.

[2] H. Bai, G. Richard, and L. Daudet. "Late reverberation synthesis : From radiance transfer to feedback delay networks." Audio, Speech, and Language Processing, IEEE/ACM Transactions on, 23(12) :2260–2271, 2015.

[3] S. Lee, H. -S. Choi and K. Lee, "Differentiable Artificial Reverberation," in *IEEE/ACM Transactions on Audio, Speech, and Language Processing*, vol. 30, pp. 2541-2556, 2022,



# Reverberation

## ■ Advantages of the perceptual approaches:

- Low complexity algorithms (implementation with IIR filters)
- Possible to define perceptual parameters that can be changed in real time.
- Ideally, a unique algorithm could simulate the full space.
- Some spaces can be very well simulated with perceptual approaches.

## ■ Drawback: not easy to link perceptual properties to specific physical properties of a room ....



# Late reverberation

- Characterised by a large density of echoes in time (*> 1000 per second in a large room*)
- Can be modelled as a Gaussian random process with exponential decay
- Similarly, the frequency response is characterised by a large density of modes (at least above a given frequency  $F_s$  called **Schroeder frequency**)
- $$F_s = 2000 \sqrt{\frac{T_{60}}{V}} \quad (T_{60} \text{ in seconds, } V \text{ in m}^3, F_s \text{ in Hz})$$





# (Late) reverberation

- Polack Model (1998)
- The reverberation (e.g. the Room Impulse Response) is modelled as a non-stationary centered Gaussian process:

$$a(t) \sim \mathcal{N}(0, r^2(t))$$

- with

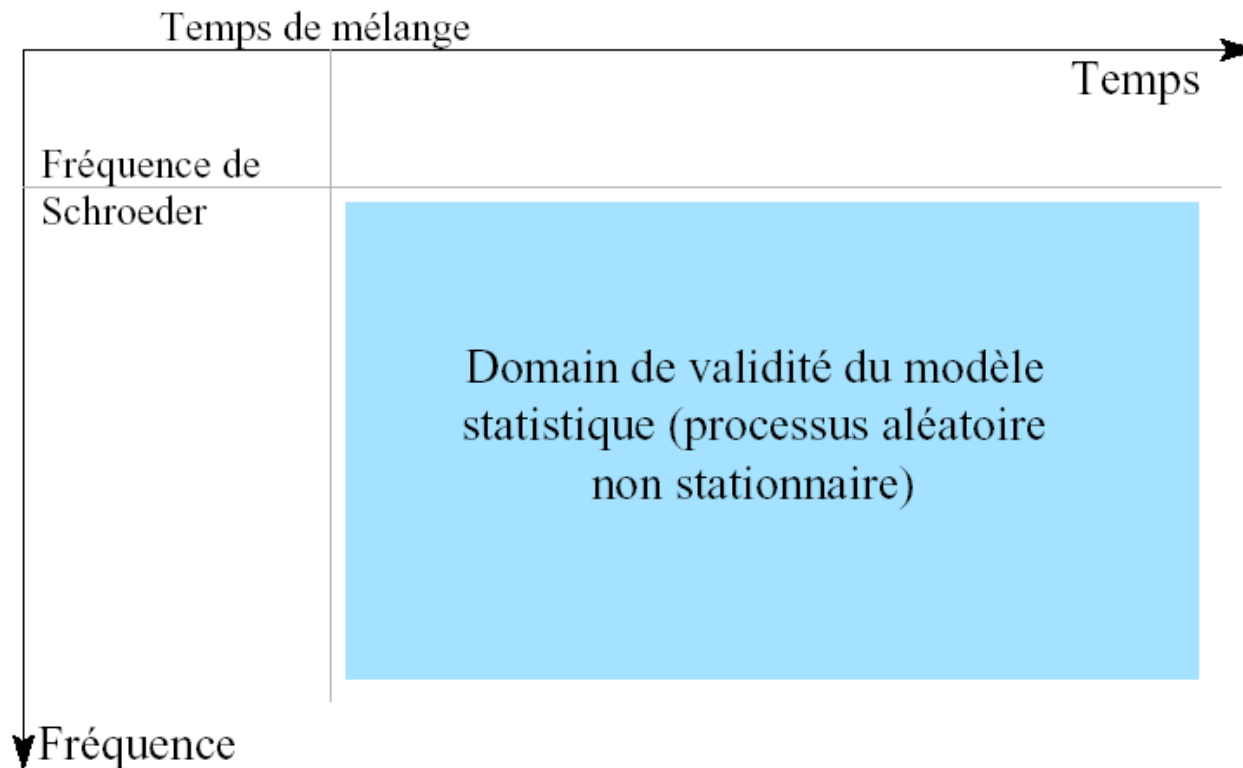
$$r^2(t) = \sigma_r^2 \exp(-2t/\tau) \quad \tau = \frac{T_{60} F_s}{3 \ln(10)} \quad \text{samples}$$

- Often used to model the full Room Impulse Response, it is theoretically (and practically) only valid for late reverberation



# Late reverberation

- Domain of validity of the statistical model: depends on the mixing time ( $\sqrt{V}$ ) and Schroeder frequency.





# Artificial reverberation algorithms

## ■ Used:

- Either to add a room effect to studio recordings
- Or to modify acoustical properties of a listening room

## ■ Early systems used analog process :

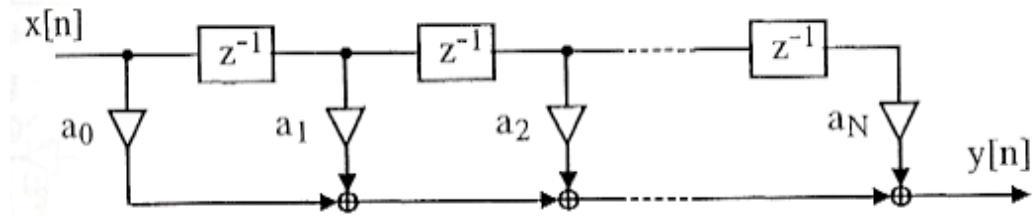
- Network of springs, metallic plates, ..

## ■ First digital systems in 1960's !!

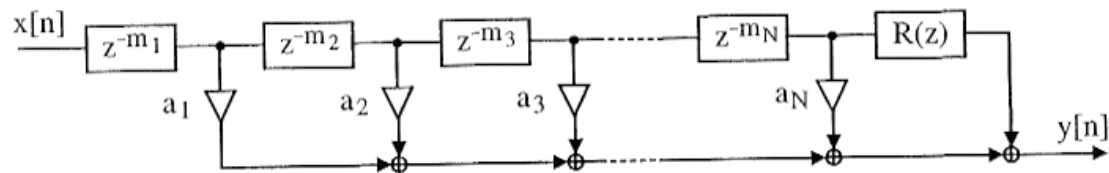


# Early echoes models

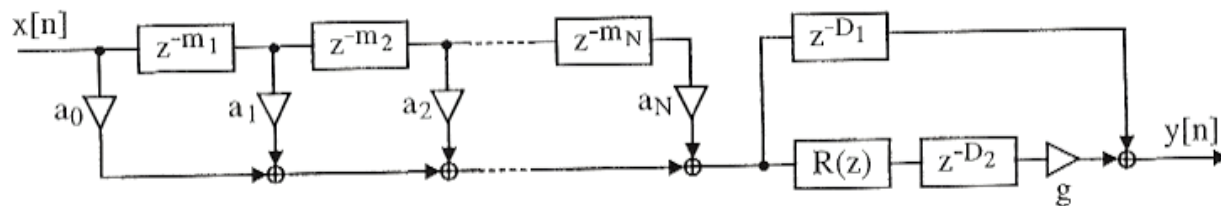
## ■ Simple model



## ■ Schroeder (1970)

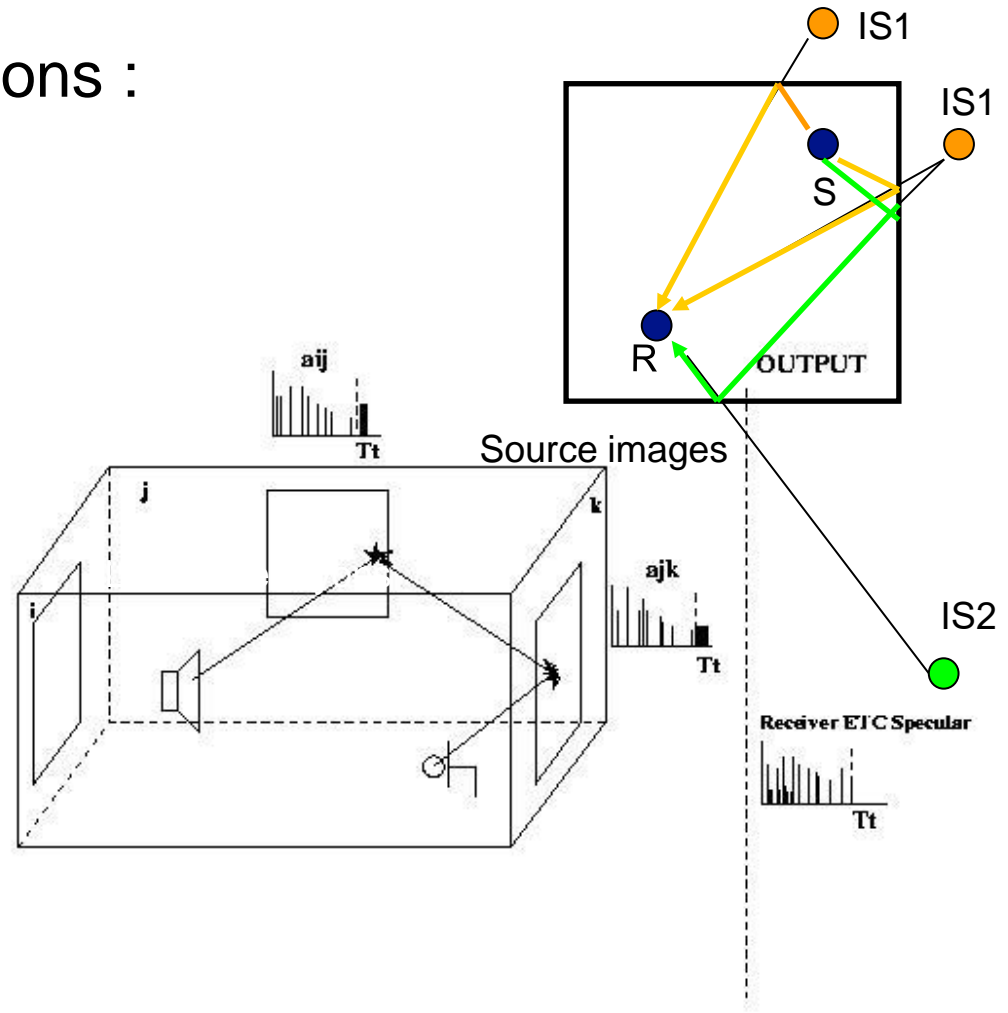


## ■ Moorer (1979)

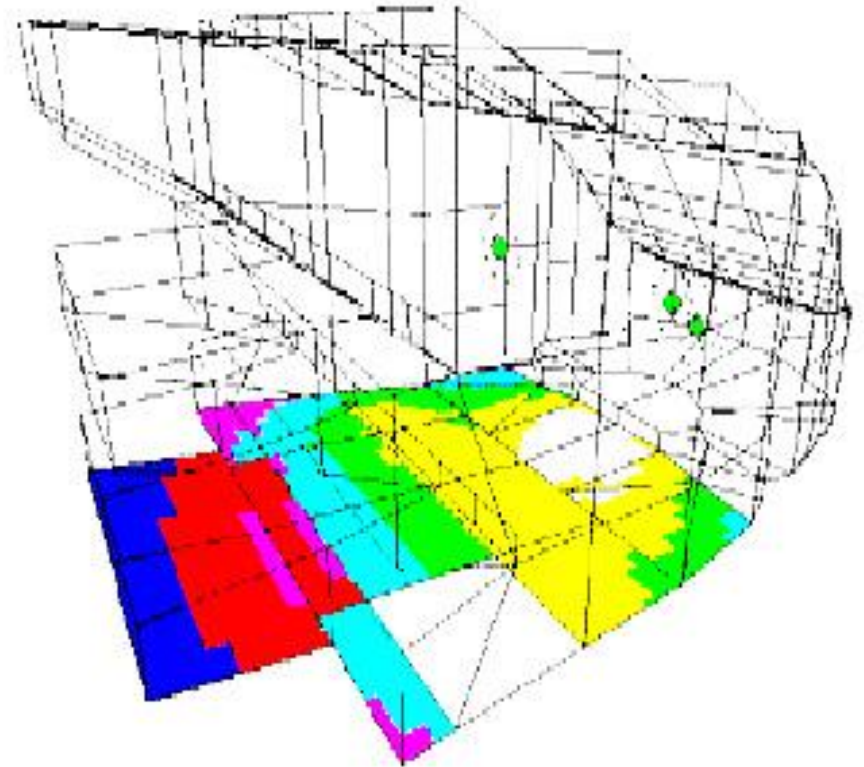


# Early echoes models

- Simulate specular reflexions :
  - Ray beams
  - « cones » beams
  - Image sources



# Application to any geometry





# Late reverberation

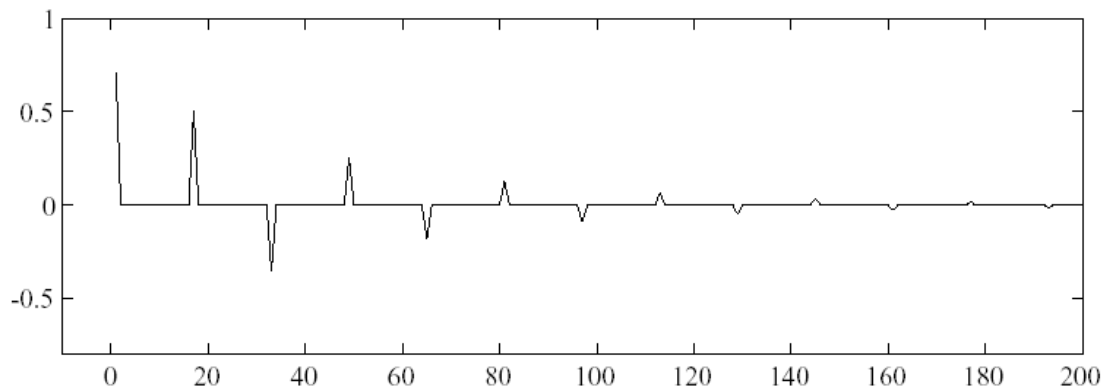
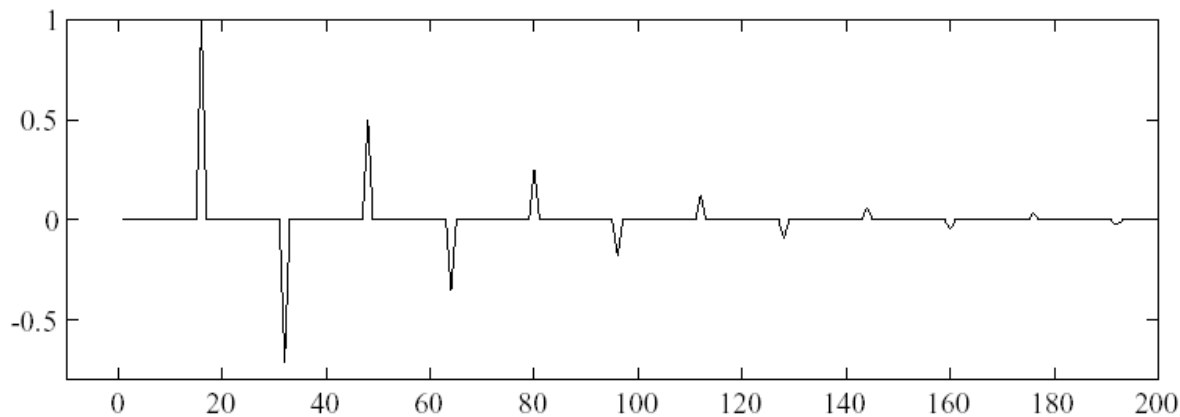
## ■ Three main difficulties :

- To obtain simultaneously a high density of modes and echoes
- To avoid a potential artificially sounding output signal (e.g. avoid the « metallic » aspect)
- To control, independently, the reverberation time and reverberated energy per frequency

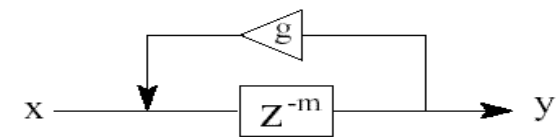


# Reverberation

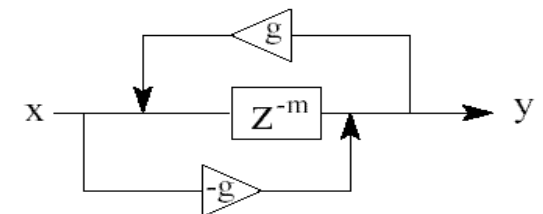
## ■ Comb filters ( $m=16, g=0.707$ )



$$C(z) = \frac{z^{-m}}{1 - gz^{-m}}$$



$$A(z) = \frac{-g + z^{-m}}{1 - gz^{-m}}$$





# Reverberation

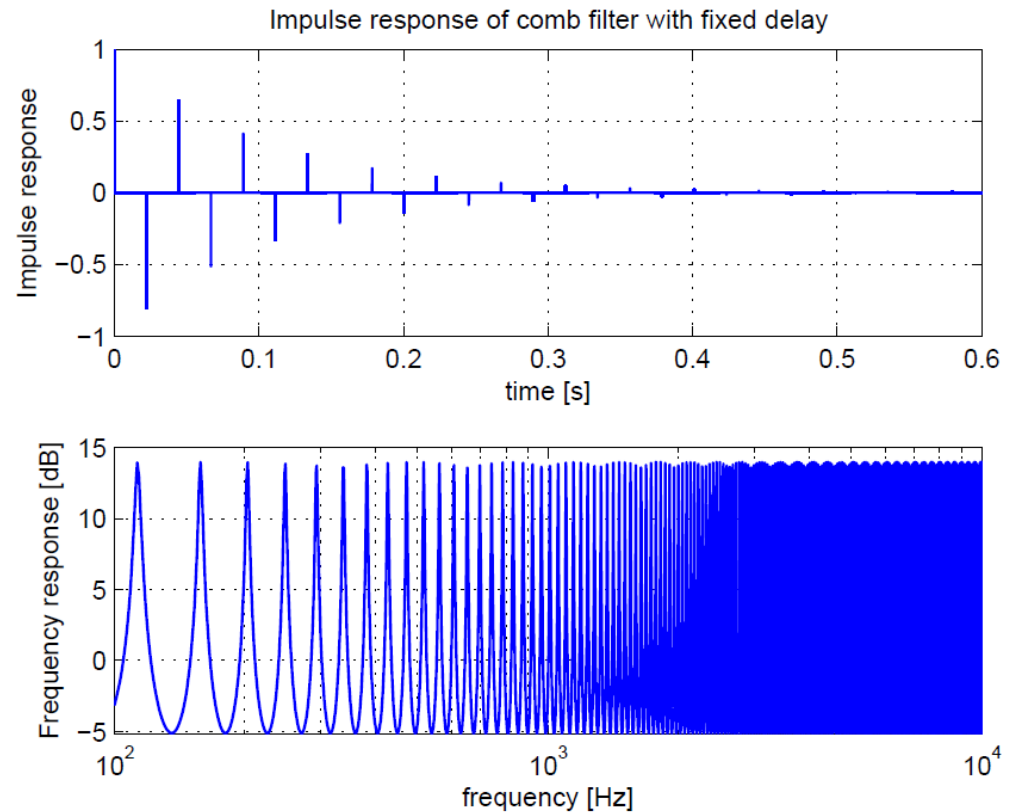
- Comb filters imply an harmonic coloration to the signal

- *Resonances at frequencies:*

$$\omega_k = \frac{2\pi k}{m}$$

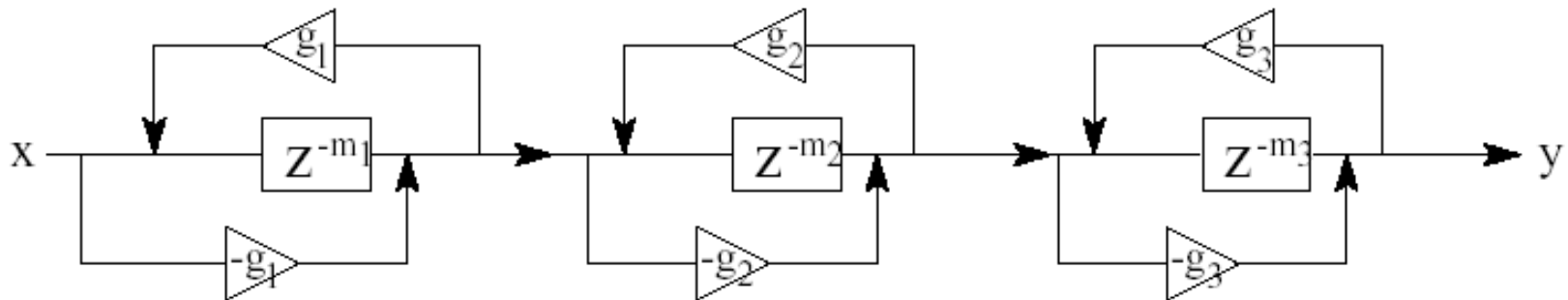
- All-pass filter also brings coloration when the signal is non-stationary (the usual case of audio signals)

- But main drawback is the low density of echoes



# Reverberation

- A solution to augment echoes density: place several all-pass filters in serie



- In practice: it induces a « metallic » sonority for transients (not very natural)



# Comb filters

- Reverberation time of a comb filter

$$\frac{20 \log_{10}(g_i)}{m_i T} = \frac{-60}{Tr}$$

- Transfer function of comb filter:

$$C(z) = \frac{z^{-m}}{1 - g z^{-m}}$$

- Modulus of the poles are given by :

$$\gamma_i = \sqrt[m_i]{g_i} = 10^{-3T/Tr}$$





# Comb filters

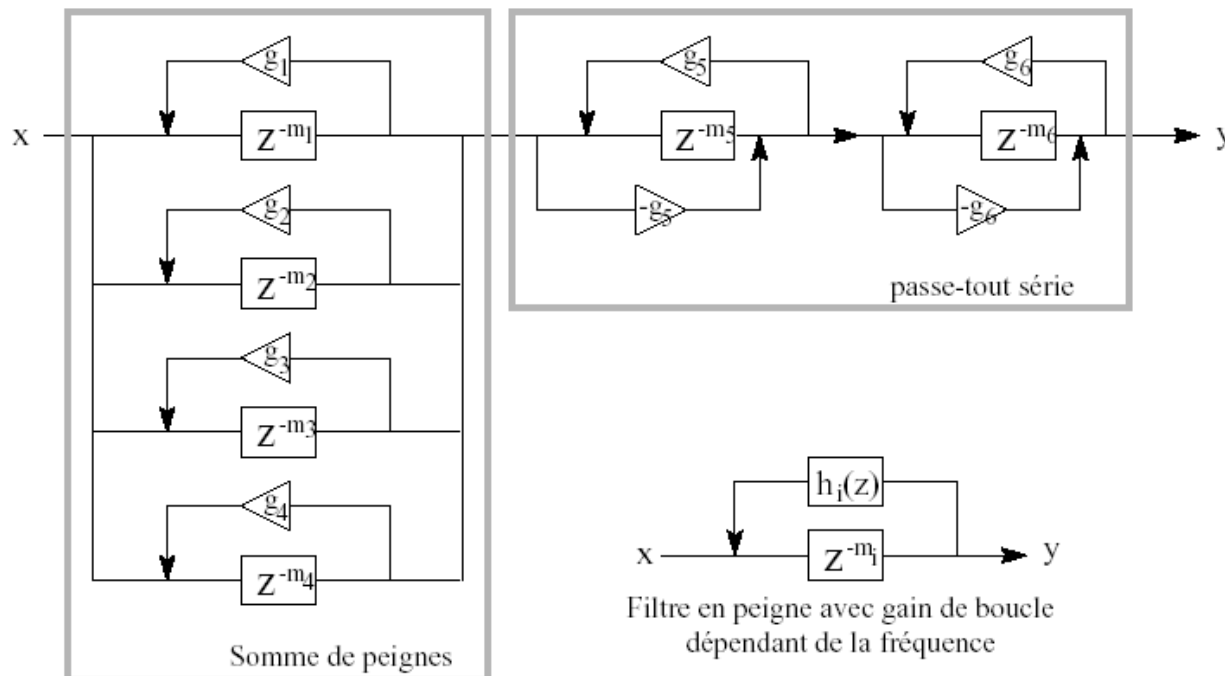
## ■ Importance of poles modulus:

- If all poles have the same magnitude, all resonances of the filters in parallel will decrease at the same speed.
- If poles have difference amplitudes, a tonal coloration may be perceived.



# Schroeder reverberator

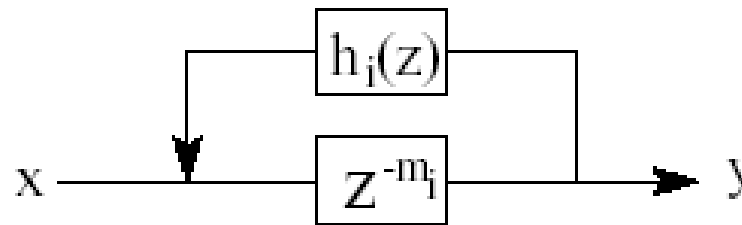
- To avoid coloration, Schroeder proposed to combine all pass and comb filters.



# Schroeder reverberator

## ■ Improvements:

- Frequency dependant reverberation time (*taking into account air absorption*)




*Comb filter with frequency dependant  
feedback gain*





# Schroeder reverberator

## ■ Example



- $Tr=0.1$  s (small room) 

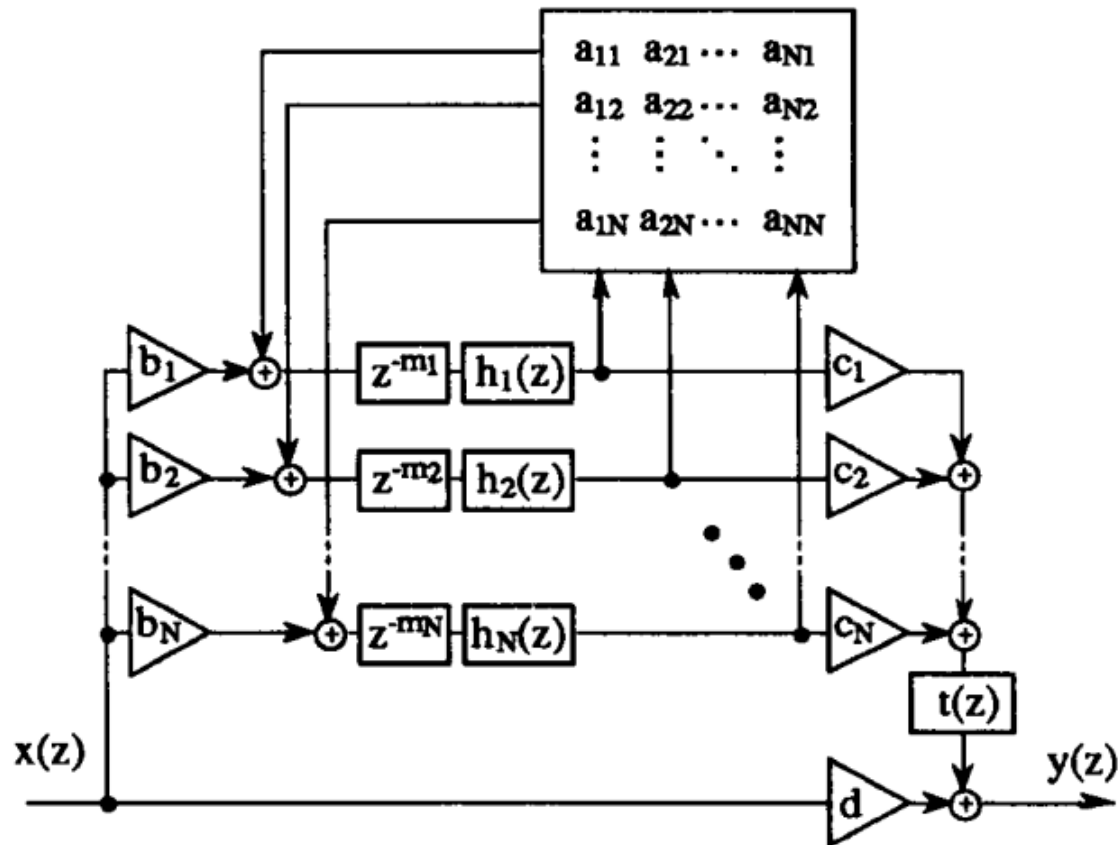
- $Tr=0.5$  s (large room) 

- $Tr=2$  s (cathedrale) 



# Reverberant filter with unitary feedback

## ■ Generalisation (*Jot et Chaigne*)







# An alternative method (physical based): Radiance Transfer Method (RTM) *(From H. Bai PhD thesis)*

## ■ Analytical acoustic radiance transfer model

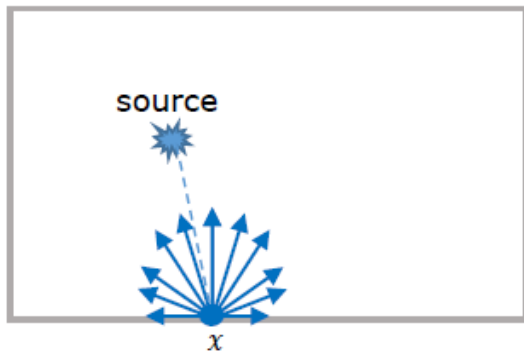
$$I(x, t) = I_0(x, t) + \int_S R(x, x', t) I(x', t - \frac{|x - x'|}{c}) dx'$$



# Une approche alternative (physique): Radiance Transfer Method (RTM)

## ■ Analytical acoustic radiance transfer model

$$I(x, t) = I_0(x, t) + \int_S R(x, x', t) I(x', t - \frac{|x - x'|}{c}) dx'$$



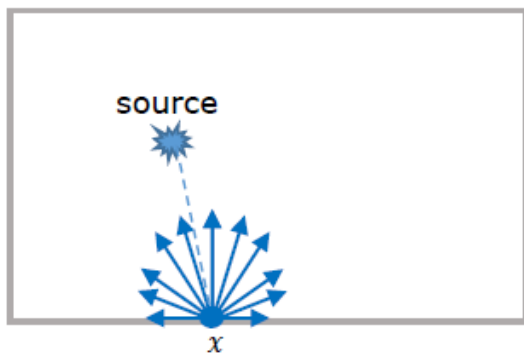
(a) Direct contribution



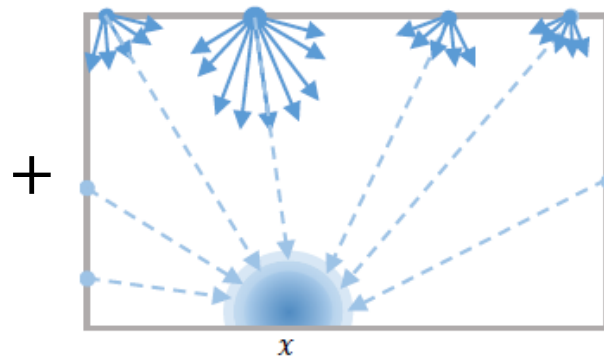
# Une approche alternative (physique): Radiance Transfer Method (RTM)

## ■ Analytical acoustic radiance transfer model

$$I(x, t) = I_0(x, t) + \int_S R(x, x', t) I(x', t - \frac{|x - x'|}{c}) dx'$$



(a) Direct contribution



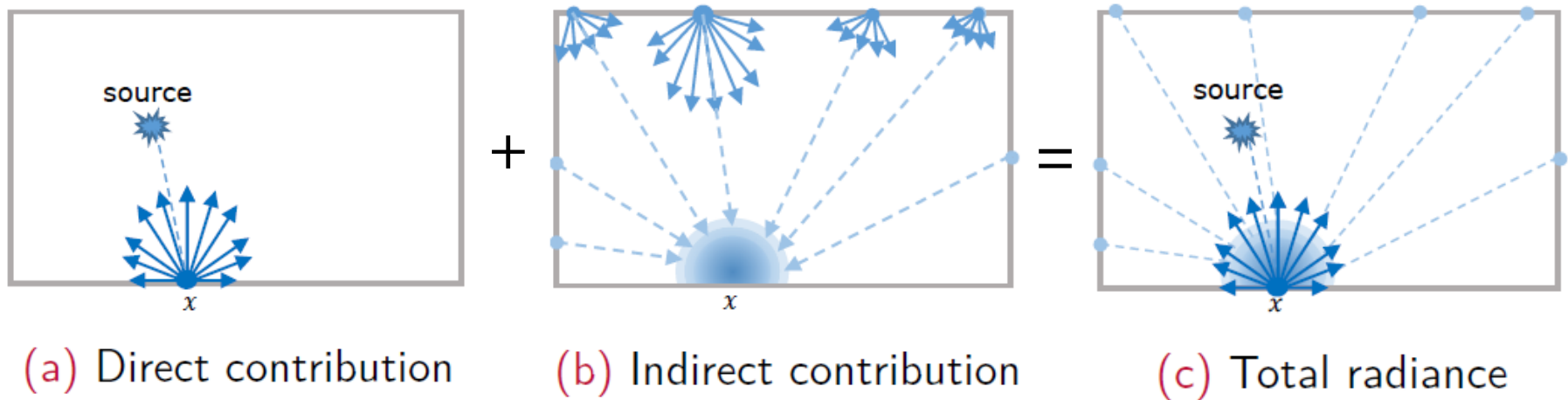
(b) Indirect contribution



# Une approche alternative (physique): Radiance Transfer Method (RTM)

## ■ Analytical acoustic radiance transfer model

$$I(x, t) = I_0(x, t) + \int_S R(x, x', t) I(x', t - \frac{|x - x'|}{c}) dx'$$

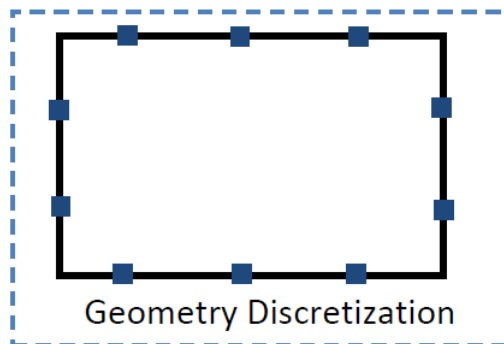
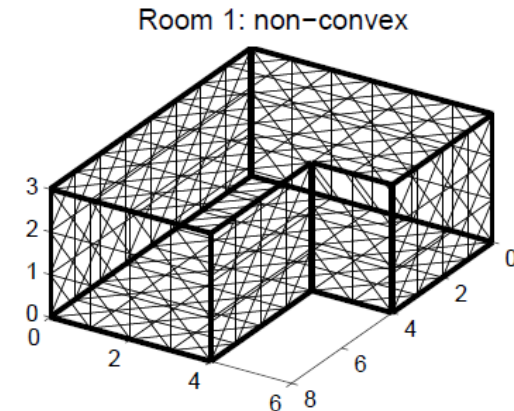


# Radiance Transfer Method: Digital simulation

## ■ Room discretization

- Room is divided in patches
- Iterative expression

$$I_i^{(n)}(t) = I_i^{(n-1)}(t) + \sum_{j=1, j \neq i}^M F_{i,j}^{(1)} I_j^{(n-1)}\left(t - \frac{r_{i,j}}{c}\right)$$

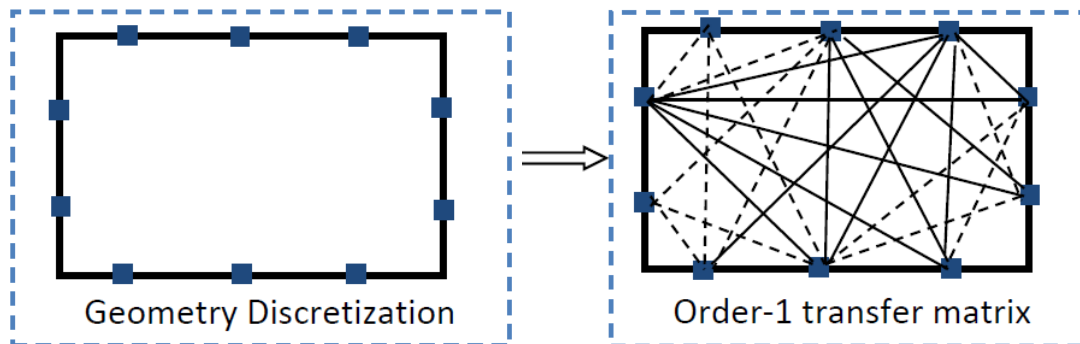
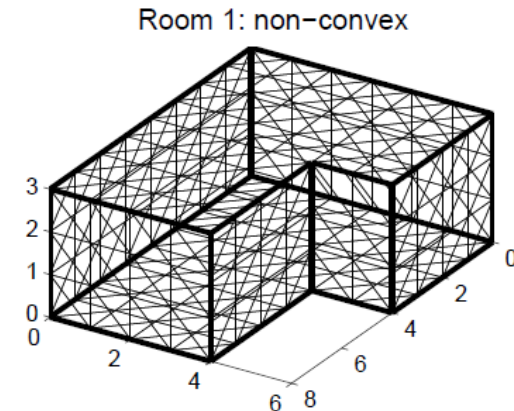


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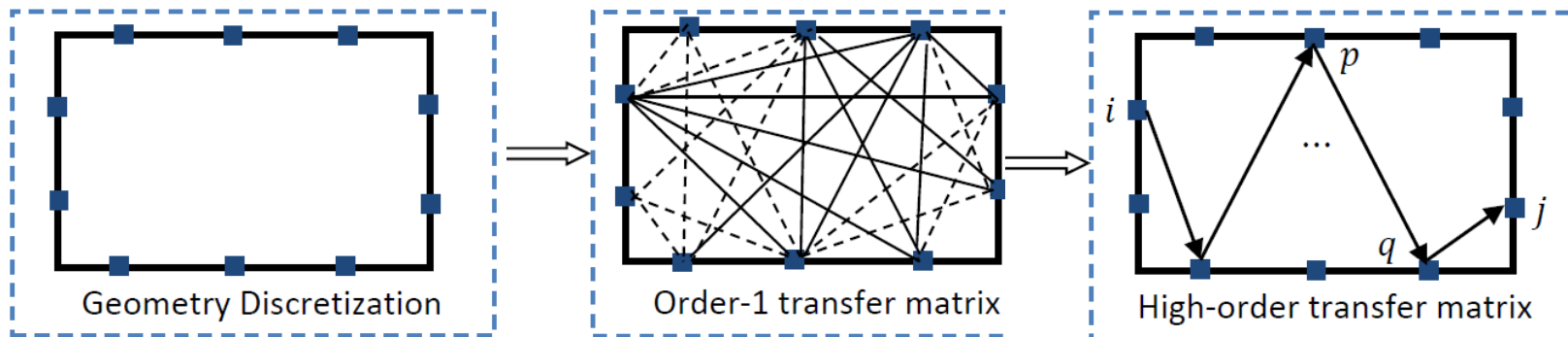
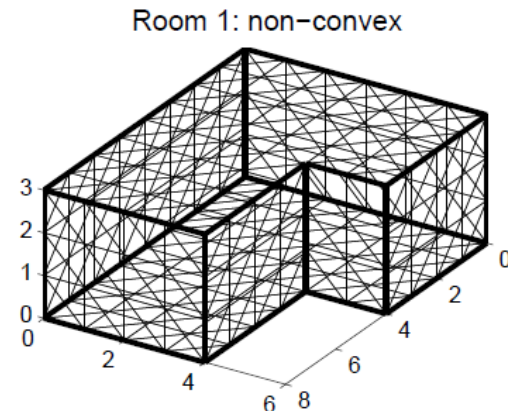


# Radiance Transfer Method: Digital simulation

## ■ Room discretization

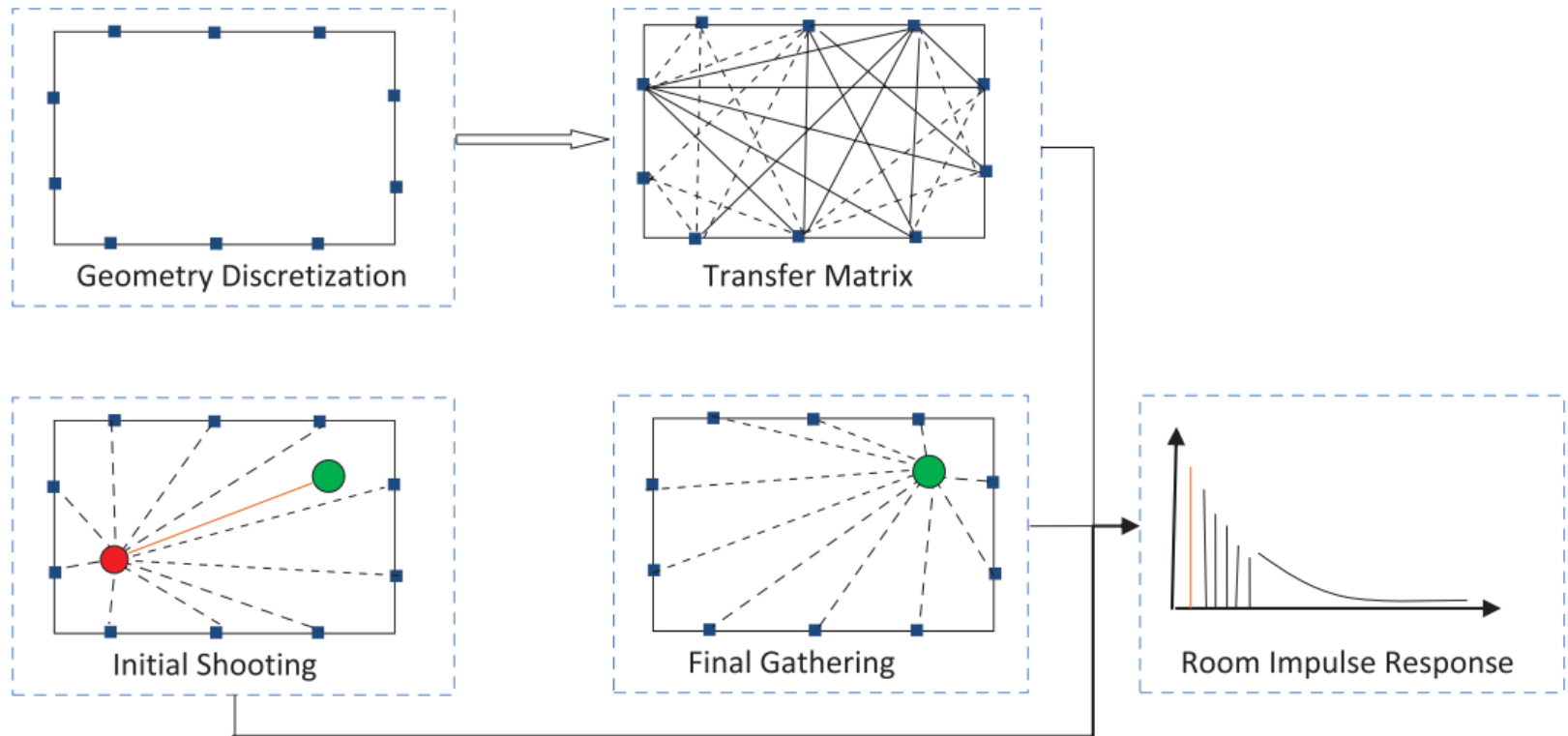
- Room is divided in patches
- Iterative expression

$$I_i^{(n)}(t) = I_i^{(0)}(t) + \sum_{j=1, j \neq i}^M F_{i,j}^{(n)} I_j^{(0)}\left(t - \frac{r_{i,j}}{c}\right)$$



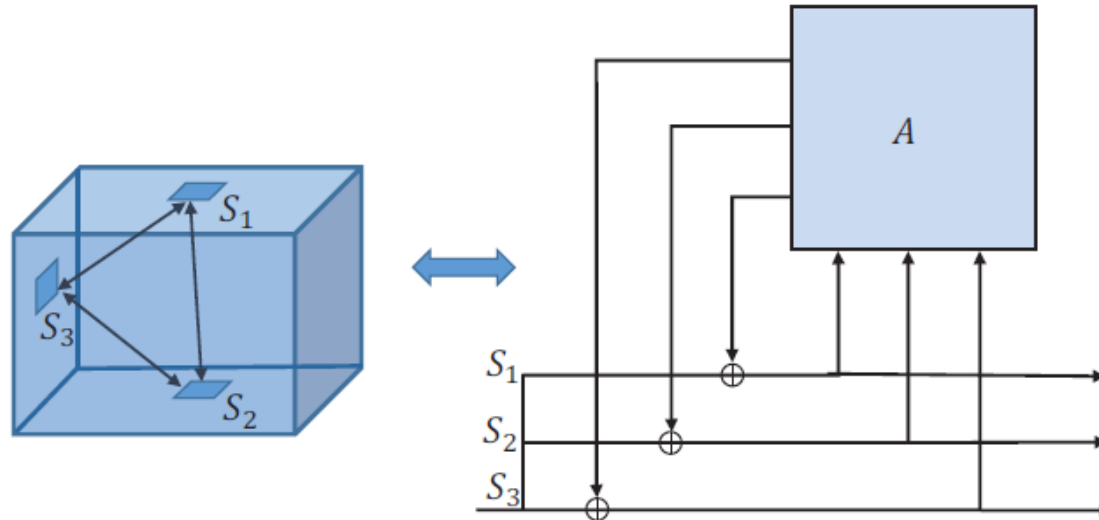
# An alternative approach: Radiance Transfer Method (RTM)

## ■ In summary





# Links between RTM and linear systems with reverberant filters



- The exchange of energy between patches of RTM can be linked to the recursive structure of the filter networks
- The exchange of energy of high order is equivalent to the infinite feedback loops of filter networks
- Brings efficient implementation of the RTM methods



# Conclusions

## ■ Solutions for artificial reverberation do exist but:

- To exactly model a reverberant space calls for complex methods even with simple systems such as with unitary feedback matrix.
- Most of the commercial systems probably used temporal variations to reduce coloration (but this is not well described in the literature).
- The perception of tonal coloration is not well understood which explains why it is not clear why a given algorithm generates natural sounding reverberation and another does not.



# A few references ....

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