# **DSP**

# **Chapter-9: Filter Bank Design**

# **Marc Moonen** Dept. E.E./ESAT-STADIUS, KU Leuven marc.moonen@esat.kuleuven.be www.esat.kuleuven.be/stadius/

# Part-IV: Filter Banks & Subband Systems

# Chapter-8

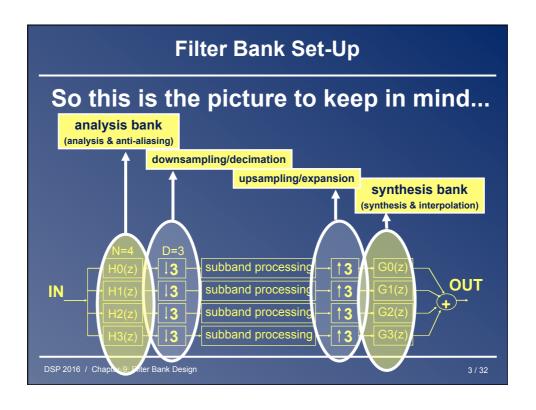
#### **Filter Bank Preliminaries**

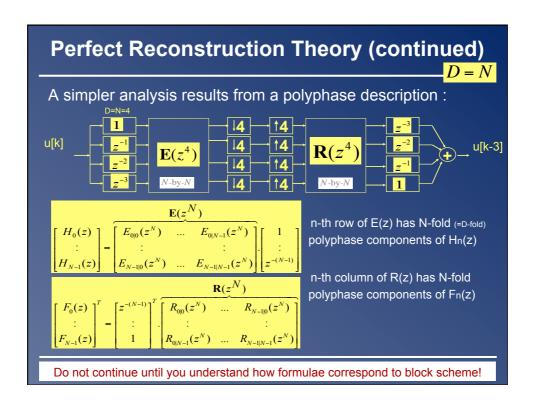
- Filter Bank Set-Up
- Filter Bank Applications
- Ideal Filter Bank Operation
- Non-Ideal Filter Banks: Perfect Reconstruction Theory

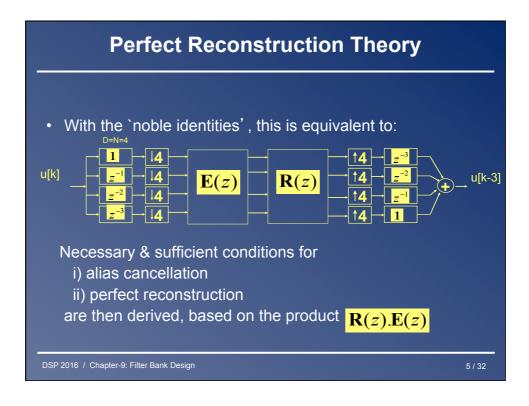
#### **Chapter-9** Filter Bank Design

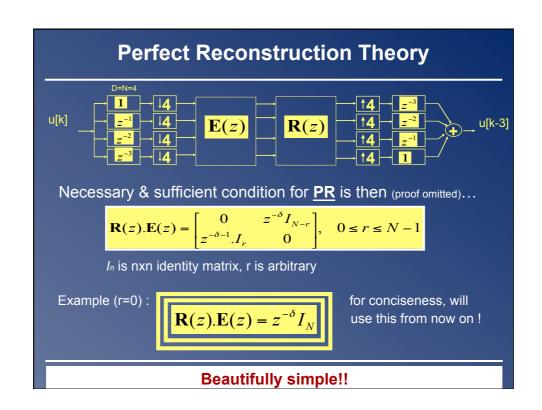
- Non-Ideal Filter Banks: Perfect Reconstruction Theory (continued)
- · Filter Bank Design Problem Statement
- General Perfect Reconstruction Filter Bank Design
- DFT-Modulated Filter Banks

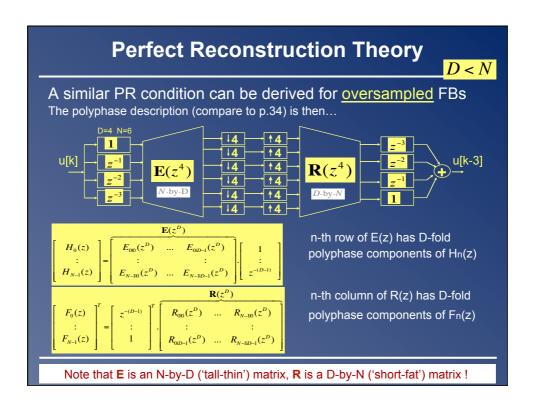
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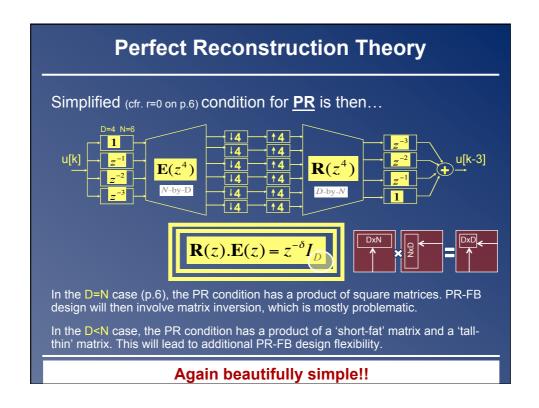












#### Filter Bank Design Problem Statement

Two design targets:

- ➡ Filter specifications, e.g. stopband attenuation, passband ripple, transition band, etc. (for each (analysis) filter!)
- **Perfect reconstruction** (PR) property.

Challenge will be in addressing two design targets at once (e.g. 'PR only' (without filter specs) is easy, see ex. Chapter-8)

PS: Can also do 'Near-Perfect Reconstruction Filter Bank Design', i.e. optimize filter specifications and at the same time minimize aliasing/distortion (=numerical optimization). Not covered here...

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#### General PR-FB Design: Maximum Decimation (D=N)

$$R(z).E(z) = z^{-\delta}I_{N}$$

(= N-by-N matrices)

- Design Procedure:
  - 1. Design all analysis filters (see Part-II).
  - 2. This determines E(z) (=polyphase matrix).
  - 3. Assuming E(z) can be inverted (?), synthesis filters are

$$\mathbf{R}(z) = z^{-\delta} \cdot \mathbf{E}^{-1}(z)$$
 (delta to make synthesis causal, see ex. p.7)

- Will consider only <u>FIR</u> analysis filters, leading to simple polyphase decompositions (see Chapter-2)
- However, FIR E(z) then generally leads to <u>IIR</u> R(z), where stability is a concern...

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# General PR-FB Design: Maximum Decimation (D=N)

PS: Inversion of matrix transfer functions ?...

- The inverse of a scalar (i.e. 1-by-1 matrix) FIR transfer function is always IIR (except for contrived examples)

$$\mathbf{E}(z) = (2 - z^{-1}) \Rightarrow \mathbf{R}(z) = \mathbf{E}^{-1}(z) = \frac{1}{(2 - z^{-1})}$$

- ...but the inverse of an N-by-N (N>1) FIR transfer function <u>can</u> be FIR

$$\mathbf{E}(z) = \begin{bmatrix} z^{-2} + \frac{1}{2} & z^{-1} \\ 2z^{-1} & 2 \end{bmatrix} \Rightarrow \mathbf{R}(z) = \mathbf{E}^{-1}(z) = \begin{bmatrix} 2 & -z^{-1} \\ -2z^{-1} & z^{-2} + \frac{1}{2} \end{bmatrix}$$
$$\det(\mathbf{E}(z)) = 1$$

PS: Compare this to inversion of integers and integer matrices

$$\mathbf{E} = 2 \Rightarrow \mathbf{R} = \mathbf{E}^{-1} = \frac{1}{2}$$
 ... but ...  $\mathbf{E} = \begin{bmatrix} 5 & 6 \\ 4 & 5 \end{bmatrix} \Rightarrow \mathbf{R} = \mathbf{E}^{-1} = \begin{bmatrix} 5 & -6 \\ -4 & 5 \end{bmatrix}$ 

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# General PR-FB Design: Maximum Decimation (D=N)

#### **Question**:

Can we build **FIR** E(z)'s (N-by-N) that have an **FIR** inverse?



#### Answer:

YES, `unimodular' E(z)'s, i.e. matrices with determinant=constant\*zd

$$\mathbf{E}(z) = \mathbf{E}_{L} \cdot \begin{bmatrix} I_{N-1} & 0 \\ 0 & z^{-1} \end{bmatrix} \cdot \mathbf{E}_{L-1} \cdot \begin{bmatrix} I_{N-1} & 0 \\ 0 & z^{-1} \end{bmatrix} \cdot \cdot \mathbf{E}_{1} \cdot \begin{bmatrix} I_{N-1} & 0 \\ 0 & z^{-1} \end{bmatrix} \cdot \mathbf{E}_{0}$$

$$\mathbf{R}(z) = \mathbf{E}_{0}^{-1} \cdot \begin{bmatrix} z^{-1} \cdot I_{N-1} & 0 \\ 0 & 1 \end{bmatrix} \cdot \mathbf{E}_{1}^{-1} \cdot \cdot \cdot \begin{bmatrix} z^{-1} \cdot I_{N-1} & 0 \\ 0 & 1 \end{bmatrix} \cdot \mathbf{E}_{L-1}^{-1} \cdot \begin{bmatrix} z^{-1} \cdot I_{N-1} & 0 \\ 0 & 1 \end{bmatrix} \cdot \mathbf{E}_{L}^{-1}$$

$$\Rightarrow \mathbf{R}(z) \cdot \mathbf{E}(z) = z^{-L} \cdot I_{N}$$

where the Ei's are constant (= not a function of z) invertible matrices

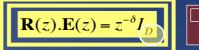
#### Design Procedure:

Optimize Ei's to meet filter specs (ripple, etc.) for <u>all</u> analysis filters (at once)

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= not-so-easy but DOABLE! 12 / 32

#### General PR-FB Design: Oversampled FBs (D<N)





- Design Procedure:
  - 1. Design all analysis filters (see Part-II).
  - 2. This determines E(z) (=polyphase matrix).
  - 3. Find R(z) such that PR condition is satisfied (how? read on...)

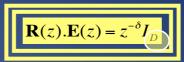
= easy if step-3 is doable...

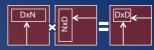
- Will consider only <u>FIR</u> analysis filters, leading to simple polyphase decompositions (see Chapter-2)
- It will turn out that when D<N an <u>FIR</u> R(z) can always be found (except in contrived cases)...

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#### General PR-FB Design: Oversampled FBs (D<N)





- Given E(z) how can R(z) be computed?
  - Assume every entry in E(z) is L<sub>E</sub>-th order FIR (i.e. L<sub>E</sub> +1 coefficients)
  - Assume every entry in R(z) is  $L_R$ -th order FIR (i.e.  $L_R$  +1 coefficients)
  - Hence number of unknown coefficients in R(z) is **D.N.(L<sub>R</sub>+1)**
  - Every entry in R(z).E(z) is (L<sub>E</sub>+L<sub>R</sub>)-th order FIR (i.e. L<sub>E</sub>+L<sub>R</sub>+1 coefficients) (cfr. polynomial multiplication / linear convolution)
  - Hence PR condition is equivalent to D.D.(L<sub>E</sub>+L<sub>R</sub>+1) linear equations in the unknown coefficients (\*)
  - Can be solved (except in contrived cases) if  $D.N.(L_R+1) \ge D.D.(L_E+L_R+1)$

(\*) Try to write down these equations!  $L_R \ge \frac{D}{(N-D)} L_E - 1$ 

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#### General PR-FB Design: Oversampled FBs (D<N)



- Given E(z) how can R(z) be computed?
  - (continued) ...
  - (continued) ...

     Can be solved (except in contrived cases) if  $L_R \ge \frac{D}{(N-D)} L_E 1$
  - If D<N, then L<sub>R</sub> can be made sufficiently large so that the (underdetermined) set of equations can be solved, i.e. an R(z) can be found (!).
  - Note that if D=N, then L<sub>R</sub> in general has to be infinitely large, i.e. R(z) in general has to be IIR

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#### **DFT-Modulated FBs**

- All design procedures so far involve monitoring of characteristics (passband ripple, stopband suppression,...) of <u>all</u> (analysis) filters, which may be tedious.
- Design complexity may be reduced through usage of `uniform' and `modulated ' filter banks.
  - DFT-modulated FBs (read on..)
  - Cosine-modulated FBs (not covered, but interesting design!)
- Will consider
  - Maximally decimated DFT-modulated FBs
  - Oversampled DFT-modulated FBs

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#### Uniform versus non-uniform (analysis) filter bank:



• N-channel uniform FB:  $H_n(z) = H_0(z \cdot e^{-j2\pi n/N})$  n = 0,...,N-1

i.e. frequency responses are <u>uniformly shifted</u> over the unit circle Ho(z)= `prototype' filter (=one and only filter that has to be designed) Time domain equivalent is:  $h_n[k] = h_0[k] . e^{j2\pi k.n/N}$ 

Non-uniform = everything that is not uniform
 e.g. for speech & audio applications (cfr. human hearing)

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H0(z)

#### Maximally Decimated DFT-Modulated FBs (D=N)

Uniform filter banks can be realized cheaply based on polyphase decompositions + DFT(FFT) (hence name `DFT-modulated FB)

#### 1. Analysis FB

If 
$$H_{0}(z), H_{1}(z), ..., H_{N-1}(z) \text{ with } H_{n}(z) = H_{0}(z.e^{-j2\pi n/N})$$

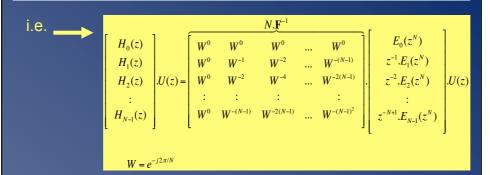
$$H_{0}(z) = \sum_{\bar{n}=0}^{N-1} z^{-\bar{n}} . E_{\bar{n}}(z^{N})$$
(N-fold polyphase decomposition)

then

$$H_{n}(z) = H_{0}(z.e^{-j2\pi n/N}) = \sum_{\bar{n}=0}^{N-1} z^{-\bar{n}}.e^{j2\pi n\bar{n}/N}.E_{\bar{n}}(z^{N} e^{-j2\pi nN/N})$$

$$= \sum_{\bar{n}=0}^{N-1} z^{-\bar{n}}.W^{-n\bar{n}}.E_{\bar{n}}(z^{N}), \quad \text{with} \quad W = e^{-j2\pi/N}$$

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#### where F is NxN DFT-matrix

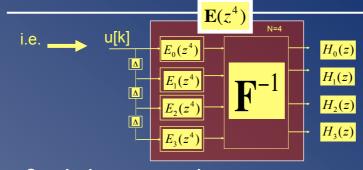
This means that filtering with the Hn's can be implemented by first filtering with the polyphase components and then applying an inverse DFT

PS: To simplify formulas the factor *N* in *N.F*<sup>-1</sup> will be left out from now on (i.e. absorbed in the polyphase components)

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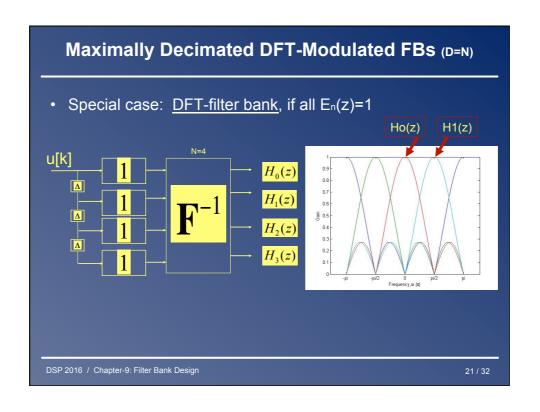
# Maximally Decimated DFT-Modulated FBs (D=N)

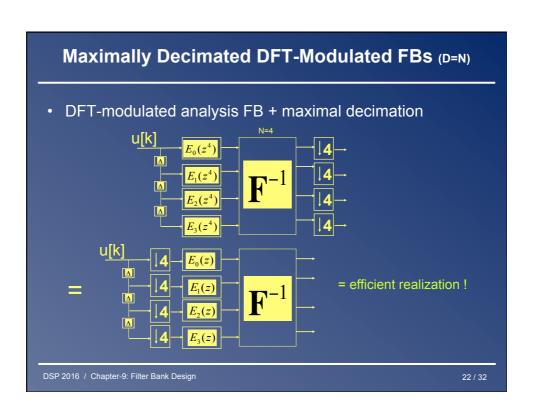


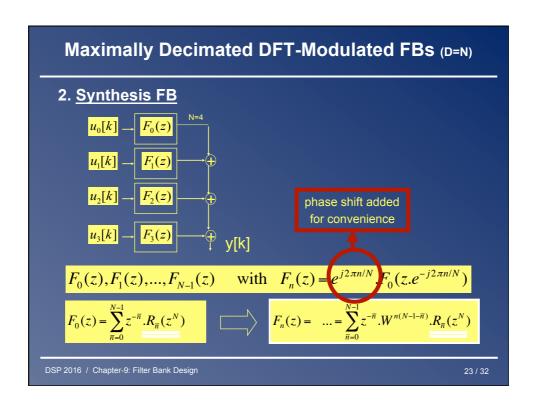
#### Conclusion: economy in...

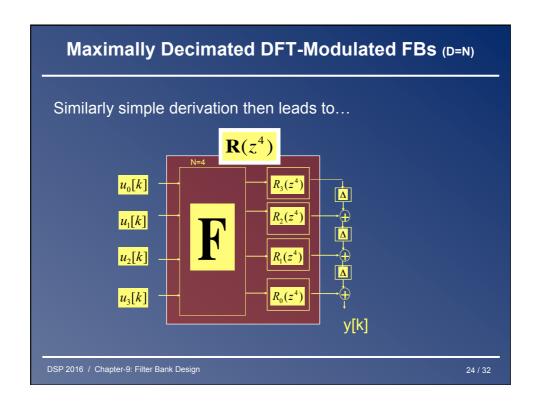
- Implementation complexity (for <u>FIR</u> filters):
   N filters for the price of 1, plus inverse DFT (=<u>FFT</u>)!
- Design complexity:
  - Design `prototype' Ho(z), then other  $H_n(z)$ 's are automatically `co-designed' (same passband ripple, etc...)!

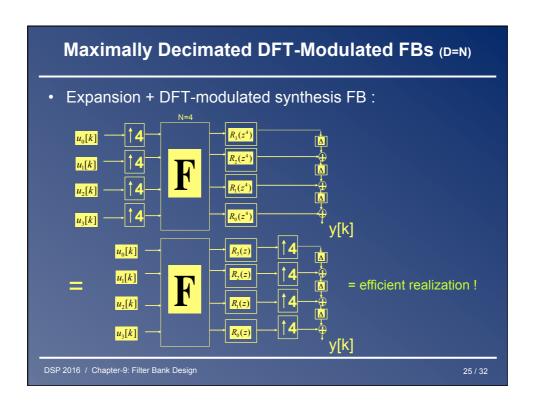
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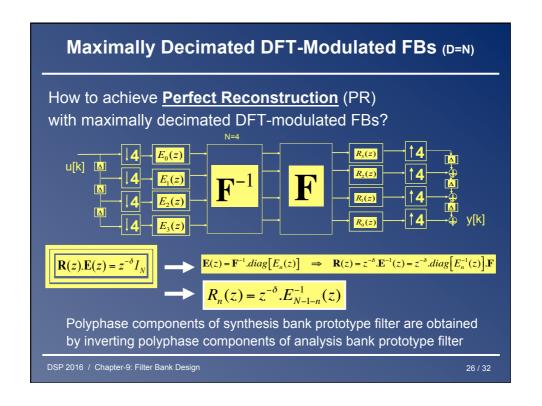














- Design Procedure:
  - 1. Design prototype analysis filter Ho(z) (see Part-II).
  - 2. This determines  $E_n(z)$  (=polyphase components).
  - 3. Assuming all E<sub>n</sub>(z)'s can be inverted (?), choose synthesis filters

$$R_n(z) = z^{-\delta} . E_{N-1-n}^{-1}(z)$$

- Will consider only <u>FIR</u> prototype analysis filter, leading to simple polyphase decomposition (Chapter-2).
- However, FIR  $E_n(z)$ 's generally again lead to IIR  $R_n(z)$ 's, where stability is a concern...

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# Maximally Decimated DFT-Modulated FBs (D=N)

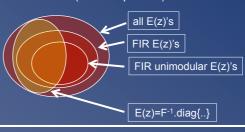
#### This does not leave much design freedom...



• FIR E(Z)? ..such that Rn(z) are also FIR

Only obtained when each  $E_n(z)$  is 'unimodular', i.e.  $E_n(z)$ =constant.z<sup>d</sup>

**Simple example** is  $E_n(z) = w_n \Rightarrow R_{N-1-n}(z) = w_n^{-1}$ , where  $w_n$ 's are constants, which leads to `windowed' IDFT/DFT bank, a.k.a. `short-time Fourier transform' (see Chapter-14)



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• Bad news: Not much design freedom for maximally decimated DFT-modulated FB's...



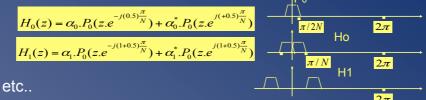
- Good news: More design freedom with...
  - Cosine-modulated FB's
  - Oversampled DFT-modulated FB's

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#### **Cosine Modulated FBs**

- Procedure:
  - Po(z) = prototype lowpass filter, cutoff at  $\frac{\pm \pi/2N}{1}$  for N filters Then...

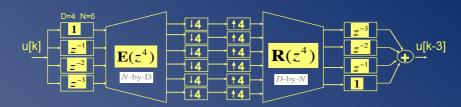


PS: Real-valued filter coefficients here!

- Details: See literature...
- Maximally decimated & oversampled FB designs
- Design software available (e.g.Matlab)

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## Oversampled DFT-Modulated FBs (D<N)



· In maximally decimated DFT-modulated FB, we had

$$\mathbf{E}(z) = \mathbf{F}^{-1}.diag[E_n(z)] \quad \mathbf{R}(z) = diag[R_{N-1-n}(z)].\mathbf{F} \quad \text{(N-by-N matrices)}$$

· In oversampled DFT-modulated FB, will have

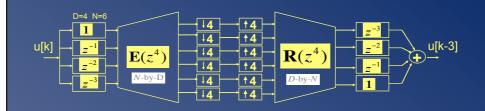
$$\underbrace{\mathbf{E}(z)}^{\text{N-by-D}} = \underbrace{\mathbf{F}^{-1}}^{\text{N-by-N}} \cdot \underbrace{\mathbf{B}(z)}^{\text{N-by-D}} = \underbrace{\mathbf{R}(z)}^{\text{D-by-N}} = \underbrace{\mathbf{C}(z)}^{\text{D-by-N}} \cdot \underbrace{\mathbf{F}}^{\text{N-by-N}}$$

with B(z) (tall-thin) and C(z) (short-fat) structured/sparse matrices constructed with poyphase components of prototype filters

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# Oversampled DFT-Modulated FBs (D<N)



- · Details: see literature
- Design Examples: http://homes.esat.kuleuven.be/~dspuser/DSP-CIS/2016-2017/material.htm
- Design software available (e.g.Matlab)

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