

# Processamento Digital de Sinais

## MÓDULO 8

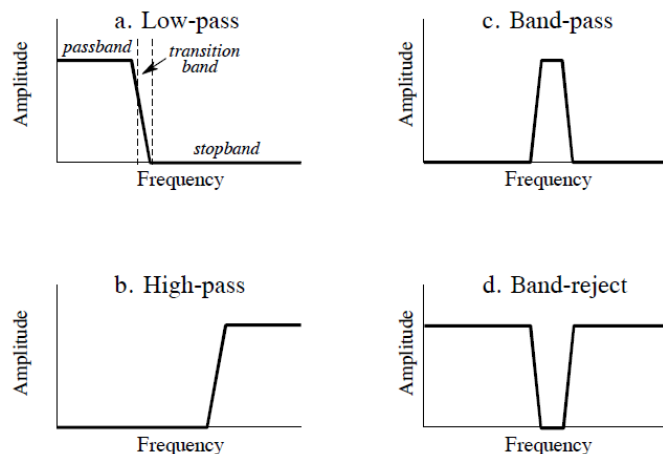
### Filtros Personalizados e Convolução por FFT

Gustavo Luís F. Vicente

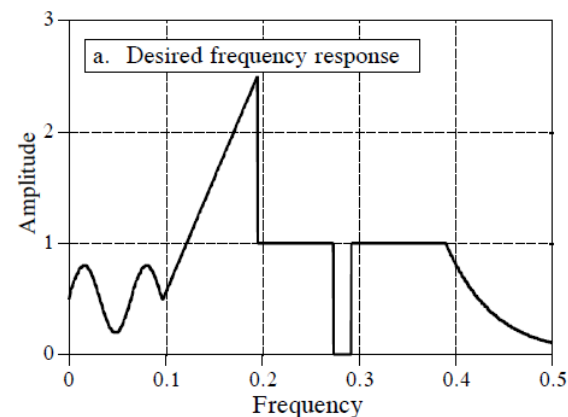
# Filtros Digitais Personalizados

## Introdução

- Vimos a implementação de filtros com comportamentos-padrão:



- E se desejarmos / precisarmos de um filtro com comportamento diferente destes?  
Por exemplo, algo assim:

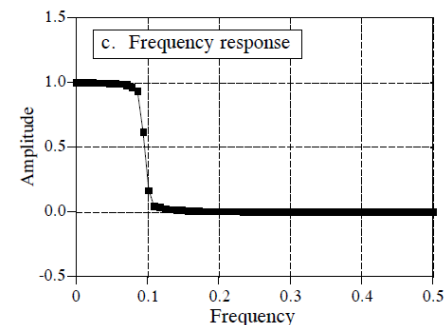
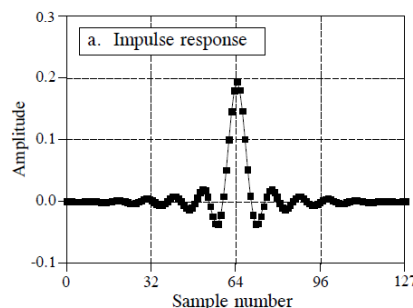


# Filtros Digitais Personalizados

## Introdução

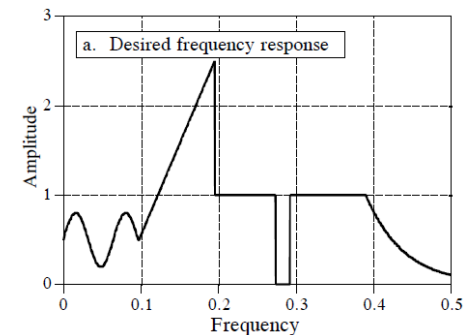
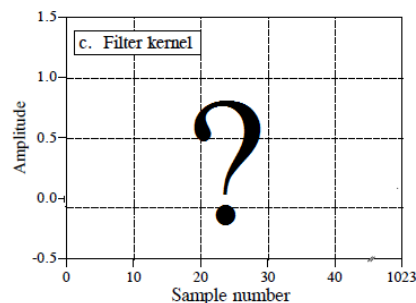
- Filtros padrão são representados por expressões matemáticas:

$$h[i] = \frac{\sin ix}{ix}$$

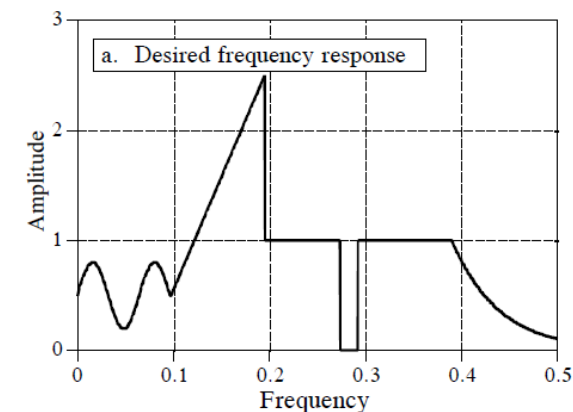


- Filtros personalizados são representados por um conjunto de valores em um vetor:

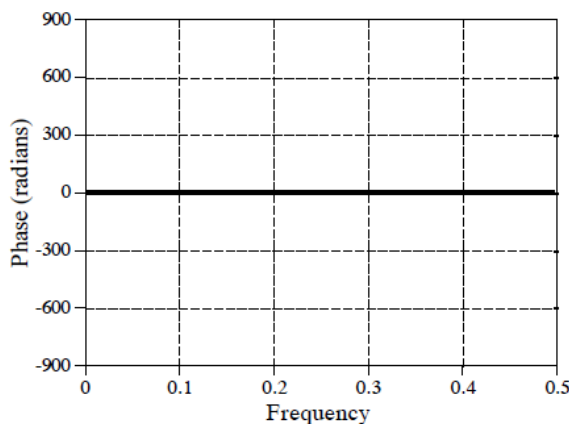
$$h[i] = \{a, b, c, \dots\}$$



# Filtros Digitais Personalizados

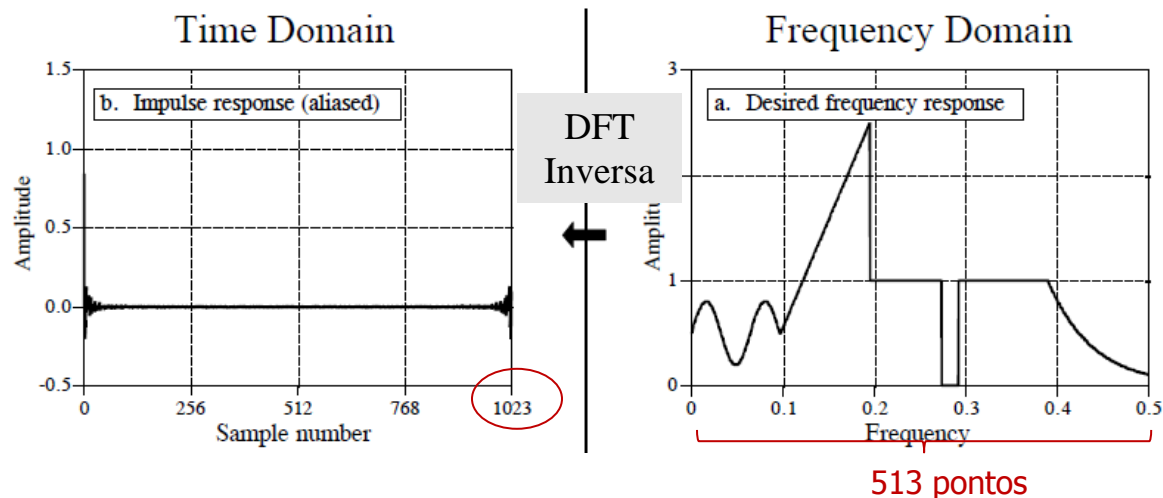


Representado por um vetor de valores que indicam a magnitude de cada componente espectral (cada frequência representada), distribuídas no intervalo entre 0 e  $0,5f_s$  (taxa de amostragem).



Deve haver um outro vetor, de mesmo tamanho, representando a fase de cada componente espectral correspondente. No nosso caso, definimos as fases com zero em todo o vetor.

# Filtros Personalizados



2000 'INVERSE FAST FOURIER TRANSFORM SUBROUTINE

2010 'Upon entry, N% contains the number of points in the IDFT, REX[ ] and

2020 'IMX[ ] contain the real and imaginary parts of the complex frequency domain.

2030 'Upon return, REX[ ] and IMX[ ] contain the complex time domain.

2040 'All signals run from 0 to N%-1.

2050 '

2060 FOR K% = 0 TO N%-1

'Change the sign of IMX[ ]

2070 IMX[K%] = -IMX[K%]

2080 NEXT K%

2090 '

2100 GOSUB 1000

'Calculate forward FFT

2110 '

2120 FOR I% = 0 TO N%-1

'Divide the time domain by N% and

2130 REX[I%] = REX[I%]/N%

'change the sign of IMX[ ]

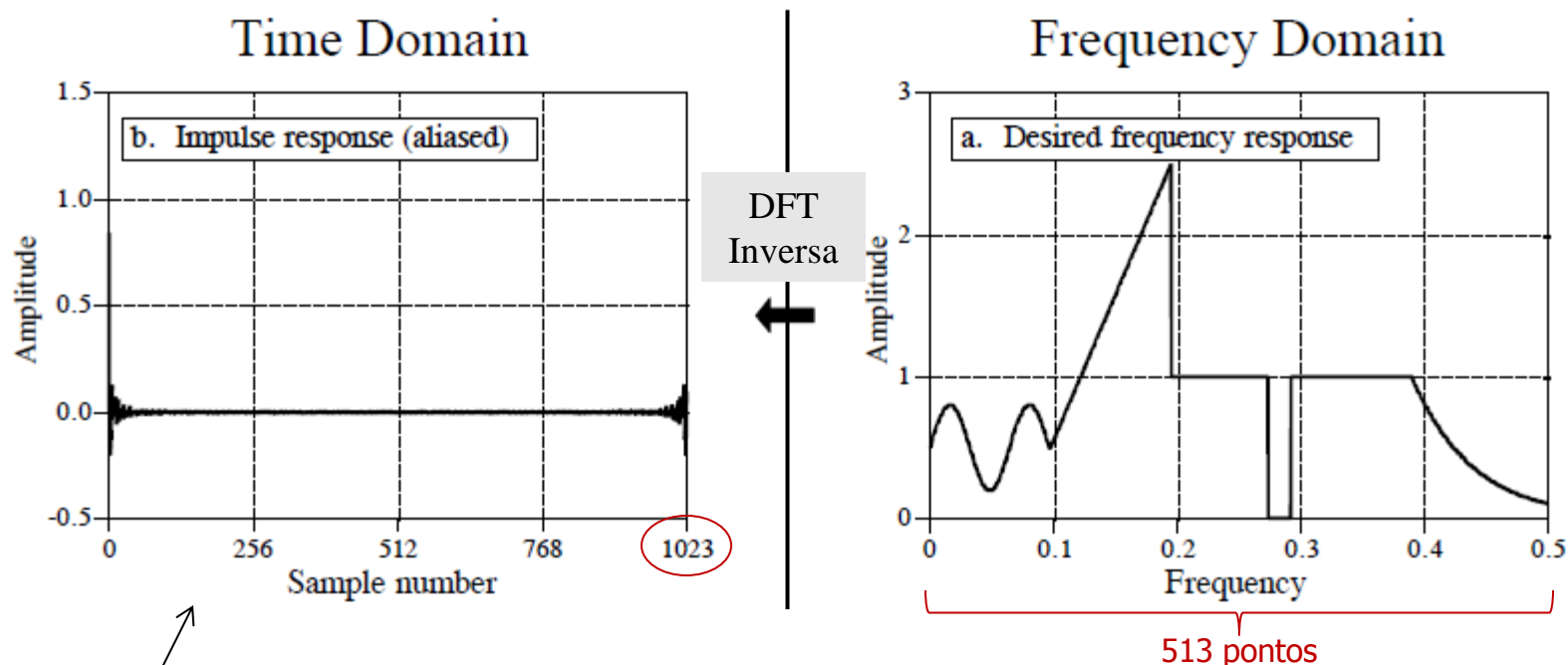
2140 IMX[I%] = -IMX[I%]/N%

2150 NEXT I%

2160 '

2170 RETURN

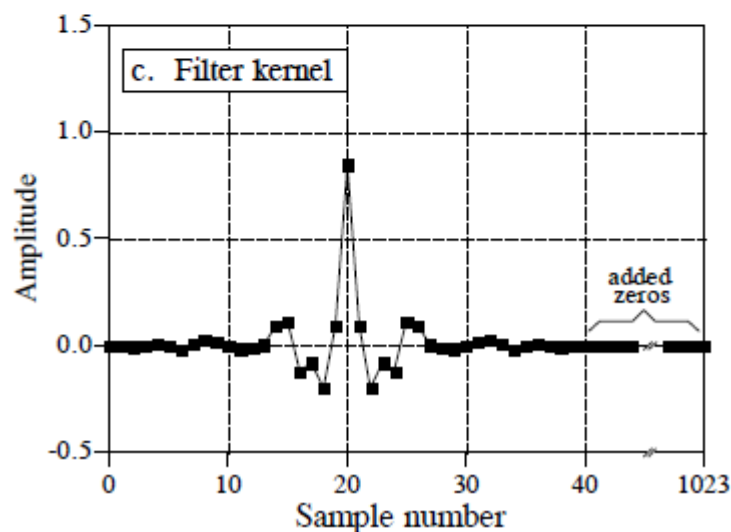
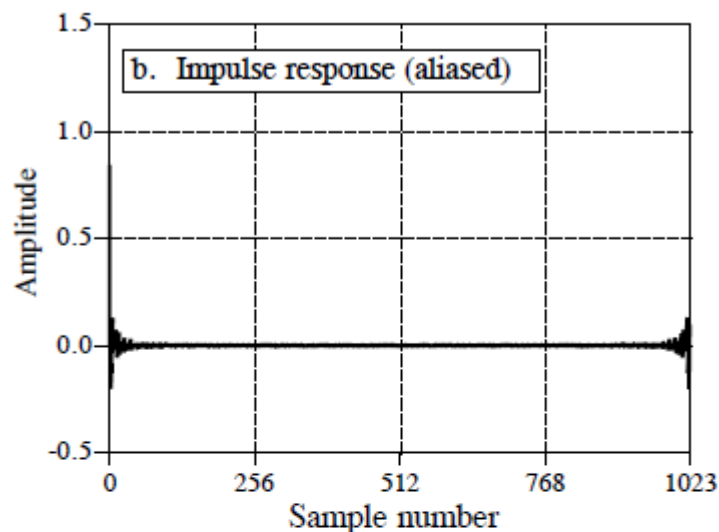
# Filtros Personalizados



- Resposta ao impulso correspondente à resposta em frequência ao lado
- Esta resposta ao impulso não está definida de forma conveniente à sua utilização em um filtro, por um computador
- Lembre-se que a DFT transforma:
  - $N$  pontos no tempo  $\rightarrow N/2+1$  pontos em frequência
  - e **vice-versa (DFT inversa)**

# Filtros Personalizados

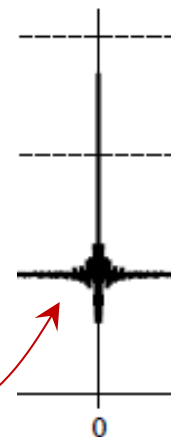
## Time Domain



- Devemos:
  - Deslocá-la  $M/2$  pontos para a direita
  - “Enjanelá-la”
  - Complementá-la com zeros à direita

Obs:

- Definimos  $M=40$ 
  - kernel do filtro com 41 pontos
- Lembremos das frequências negativas na representação de resposta ao impulso



# Filtros Personalizados

100 'CUSTOM FILTER DESIGN

110 'This program converts an aliased 1024 point impulse response into an M+1 point

120 'filter kernel (such as Fig. 17-1b being converted into Fig. 17-1c)

130 '  
140 DIM REX[1023] 'REX[ ] holds the signal being converted

150 DIM T[1023] 'T[ ] is a temporary storage buffer

160 '  
170 PI = 3.14159265  
180 M% = 40 'Set filter kernel length (41 total points)

190 '  
200 GOSUB XXXX 'Mythical subroutine to load REX[ ] with impulse response

210 '  
220 FOR I% = 0 TO 1023 'Shift (rotate) the signal M/2 points to the right

230 INDEX% = I% + M%/2

240 IF INDEX% > 1023 THEN INDEX% = INDEX%-1024

250 T[INDEX%] = REX[I%]

260 NEXT I%

270 '  
280 FOR I% = 0 TO 1023

290 REX[I%] = T[I%]

300 NEXT I%

310 ' 'Truncate and window the signal

320 FOR I% = 0 TO 1023

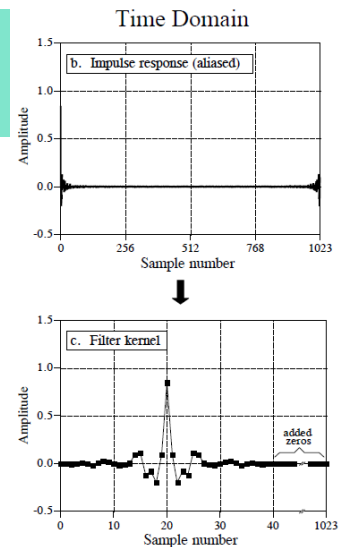
330 IF I% <= M% THEN REX[I%] = REX[I%] \* (0.54 - 0.46 \* COS(2\*PI\*I%/M%))

340 IF I% > M% THEN REX[I%] = 0

350 NEXT I%

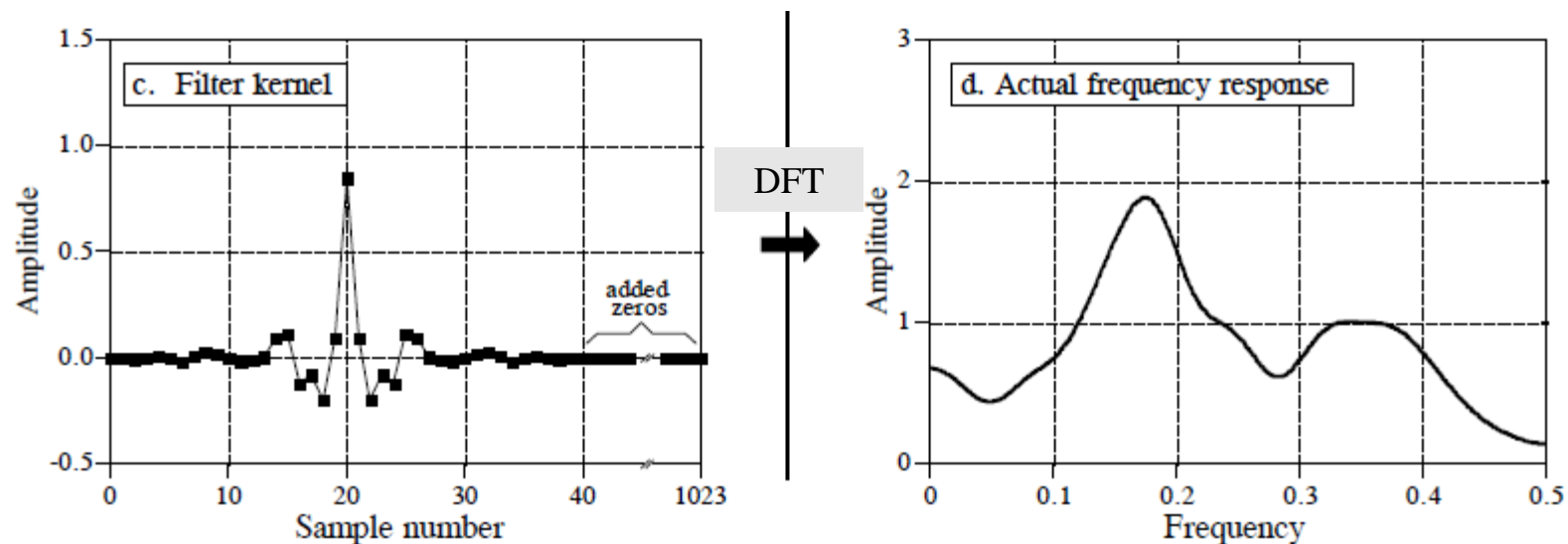
360 ' 'The filter kernel now resides in REX[0] to REX[40]

370 END





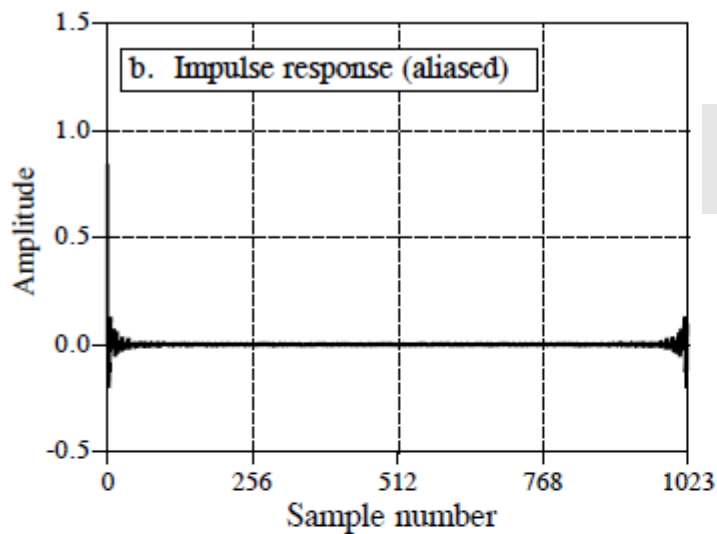
# Filtros Personalizados



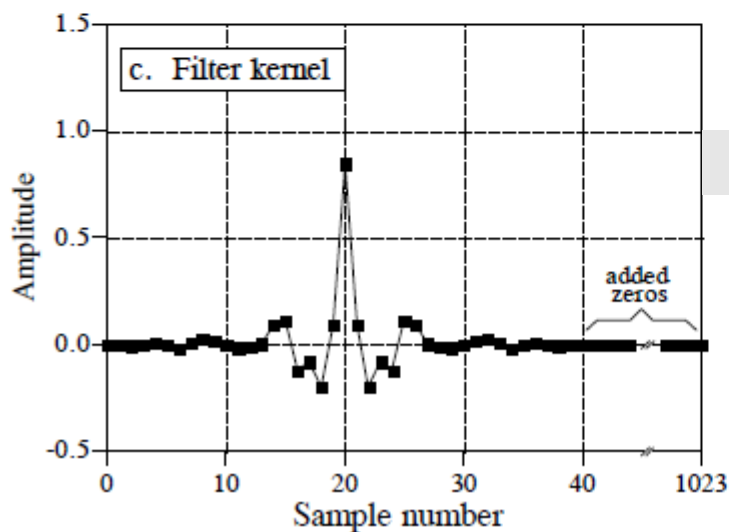
- Testamos o kernel criado, aplicando a DFT e verificando a resposta em frequência obtida

# Filtros Personalizados

Time Domain



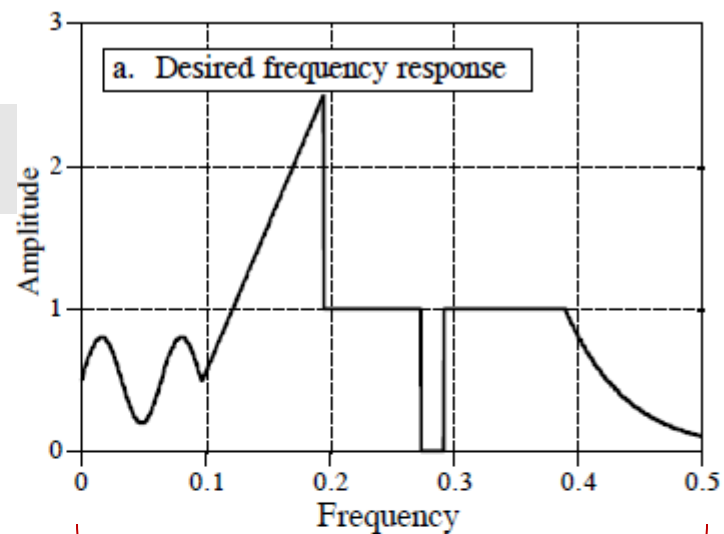
deslocamento e janelamento ↓ completa com zeros



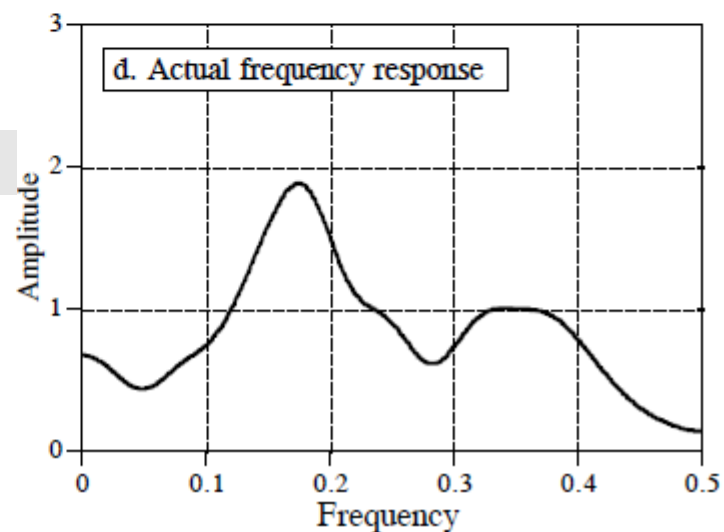
FFT  
inversa



Frequency Domain



representado por vetor de 513 elementos

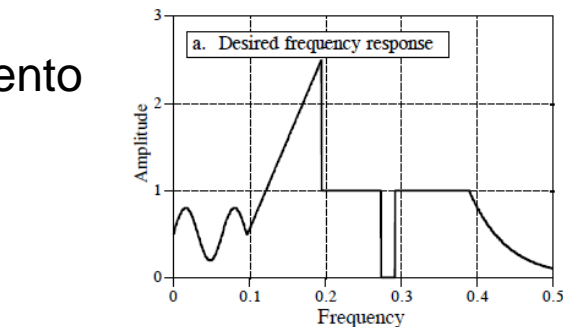
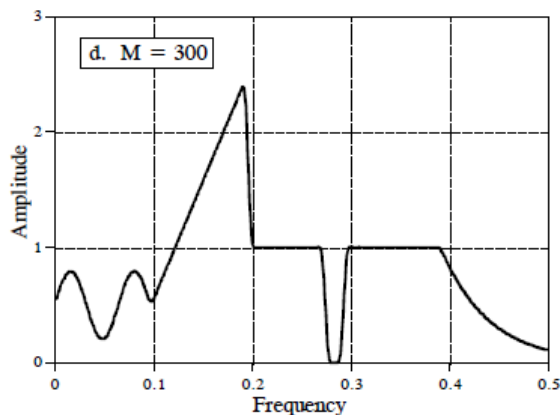
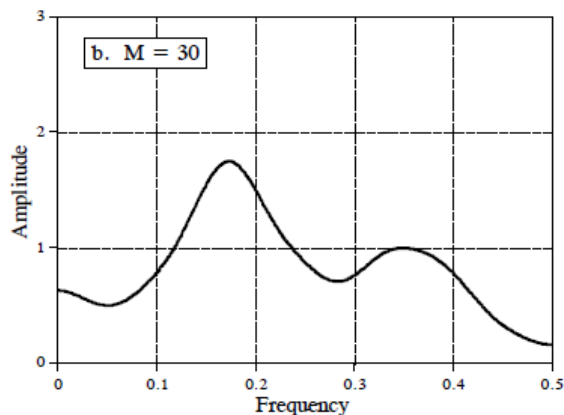
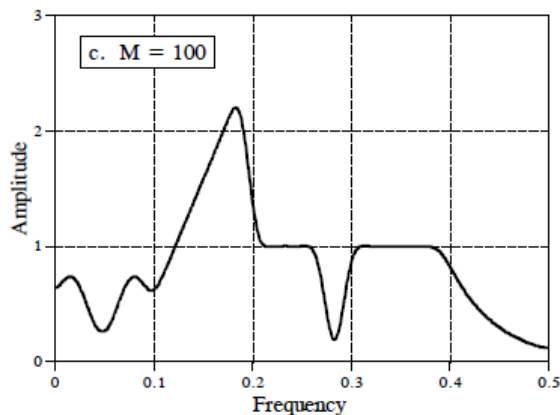
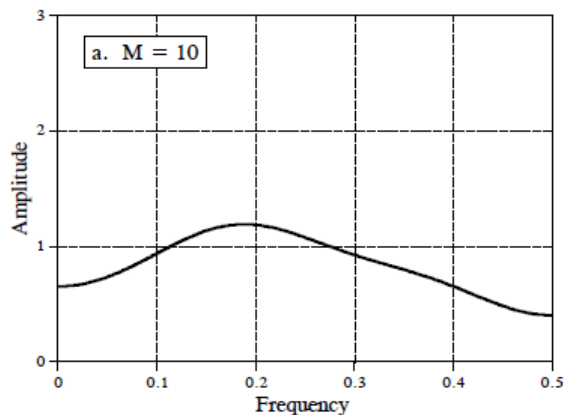


FFT



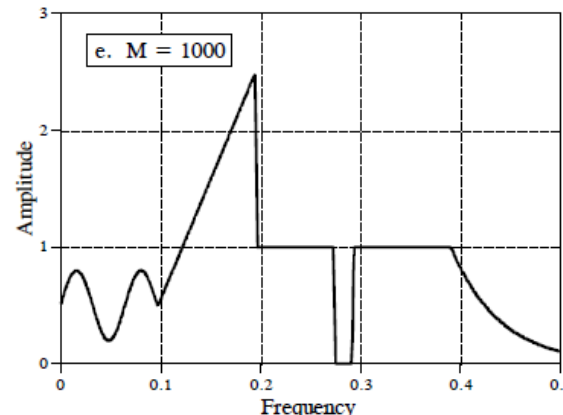
# Filtros Personalizados

- Resposta em frequência do filtro em função do tamanho do kernel ( $M$ )
- Maior  $M \rightarrow$  maior precisão  $\rightarrow$  maior processamento



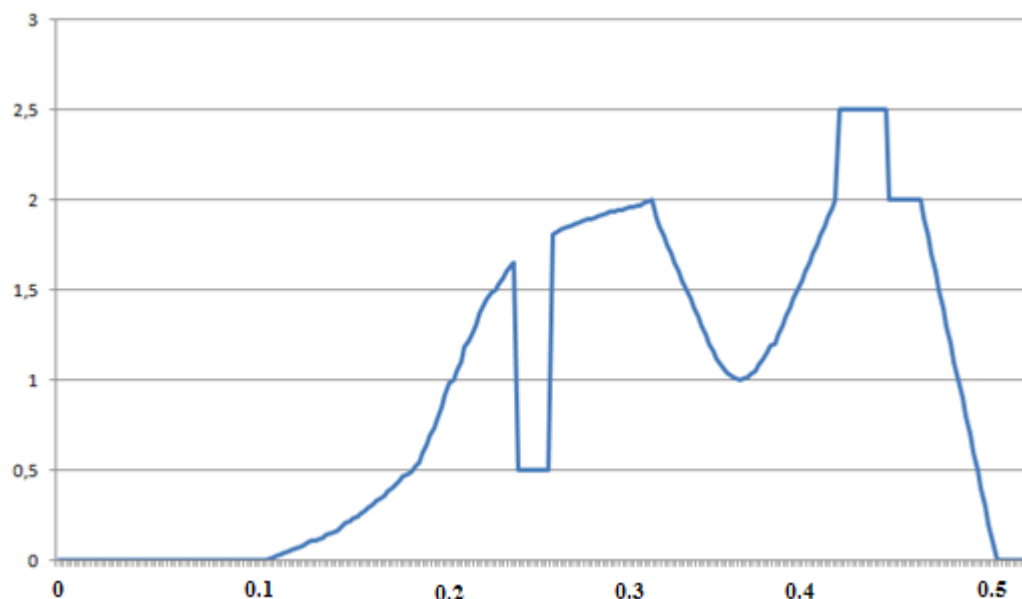
IDEAL

← REAIS  
↙  
↓



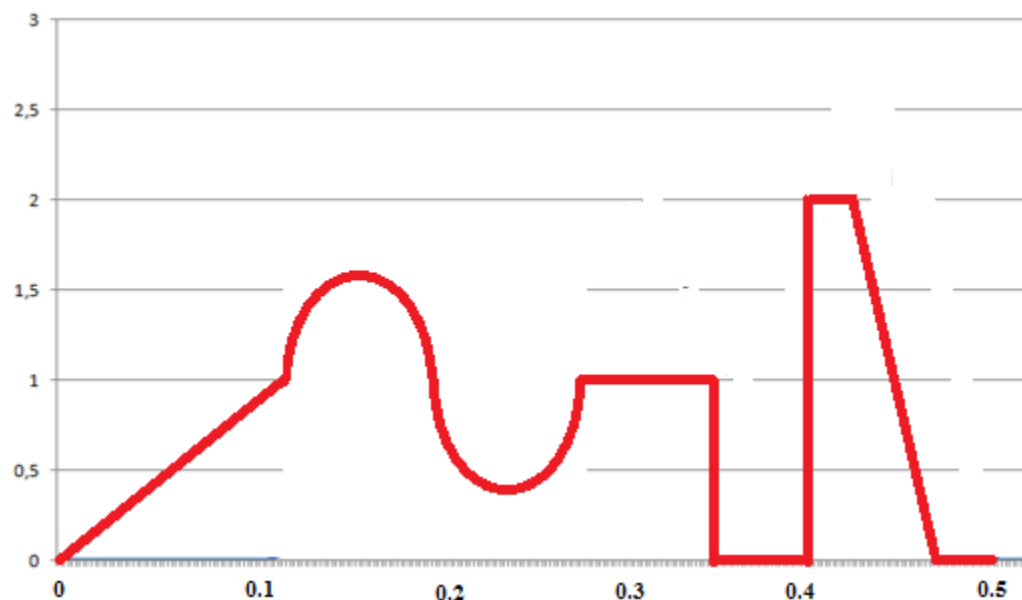
# Filtros Personalizados

- TRABALHO - Grupo 1
- Gerar o kernel do filtro que produza a resposta em frequência mostrada na figura abaixo. Gerar o kernel com os tamanhos ( $M$ ) iguais a 20, 100 e 500, testando-os e plotando as respostas em frequência correspondentes.
- Entregar no dia 23/06 (comporá a Nota2):
  - memória de cálculo resumida e gráficos



# Filtros Personalizados

- TRABALHO - Grupo 2
- Gerar o kernel do filtro que produza a resposta em frequência mostrada na figura abaixo. Gerar o kernel com os tamanhos ( $M$ ) iguais a 20, 100 e 500, testando-os e plotando as respostas em frequência correspondentes.
- Entregar no dia da prova (comporá a Nota2):
  - memória de cálculo resumida e gráficos

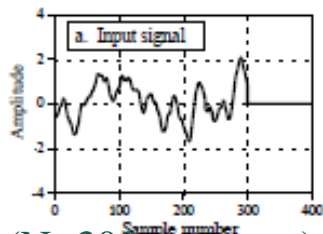


# *Método de Sobreposição*

## Situações

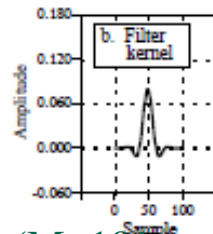
- Sinal com um conjunto de amostras muito grande
  - HiFi audio: 5Mbytes/min
  - Video: 500Mbytes/min
- Sinal processado em tempo real
  - Telefonia
- Melhora no tempo de processamento
  - Convolução por FFT
    - $a(t) * b(t) \longleftrightarrow A[f] \times B[f]$
- Lembrando que:
  - $x_N * h_M \rightarrow y_{N+M-1}$

# Método de Sobreposição



(N=300 amostras)

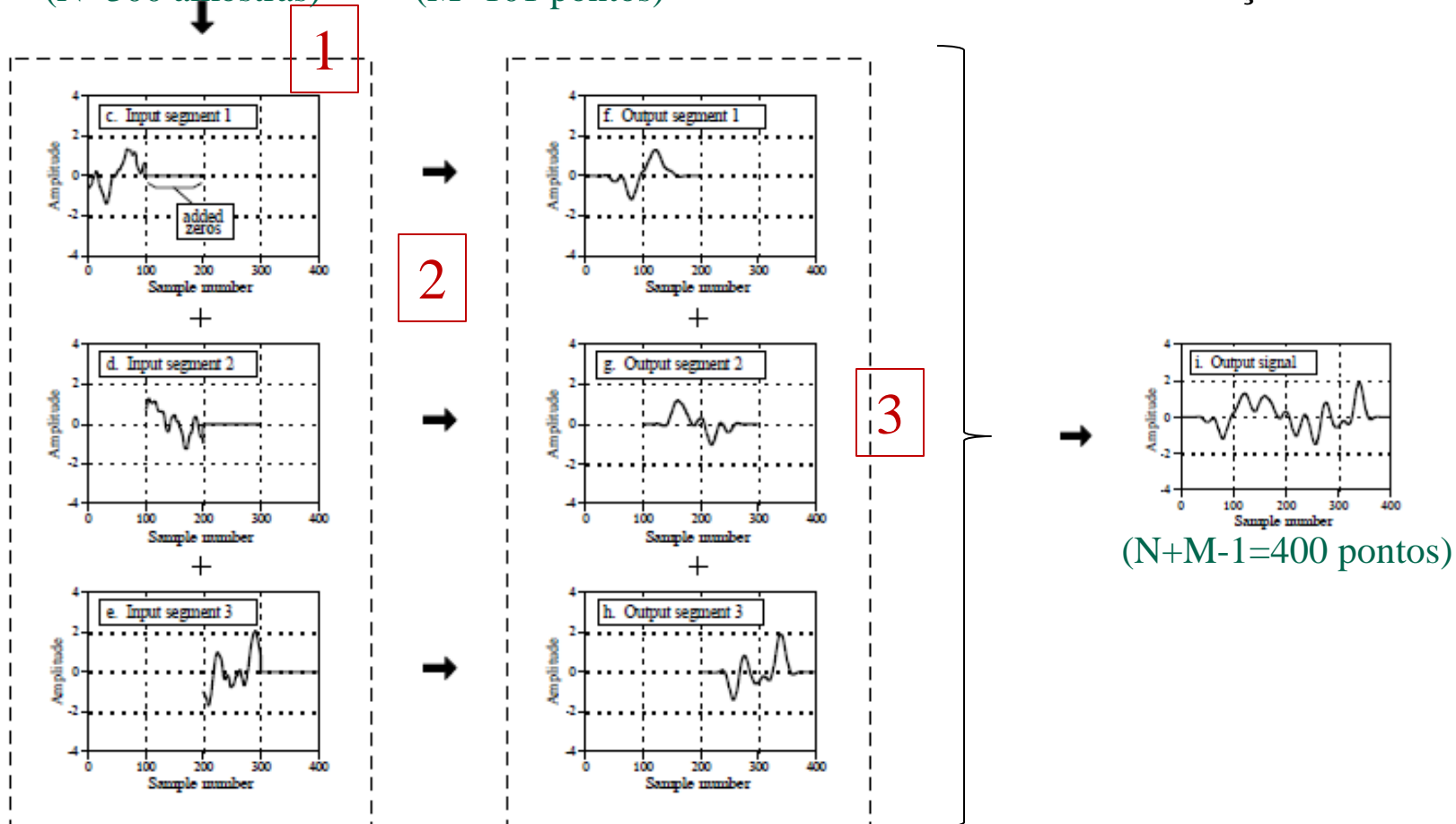
\*



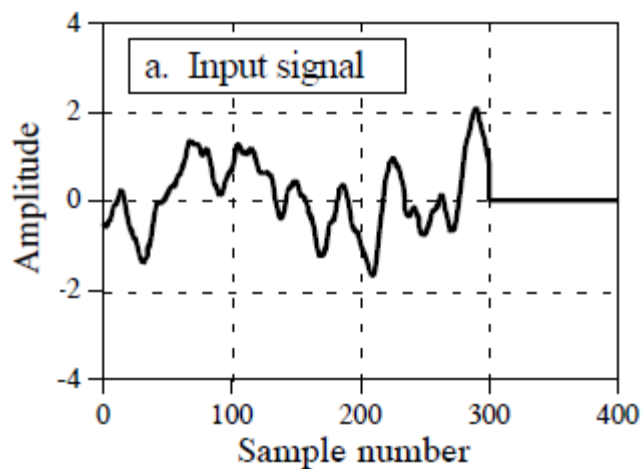
(M=101 pontos)

= ?

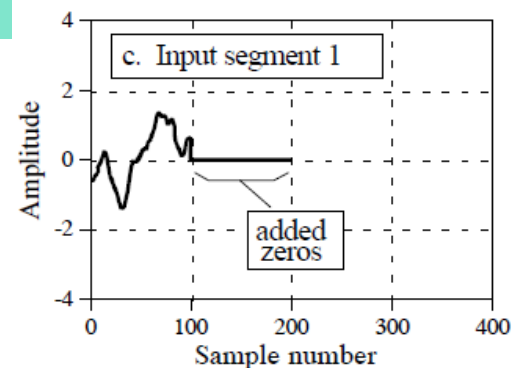
- 1) Desmembra o sinal de entrada
- 2) Convolui cada pedaço com  $h[i]$
- 3) Soma cada resultado das convoluções



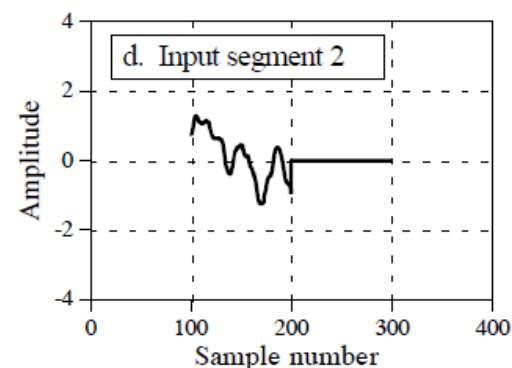
# Método de Sobreposição



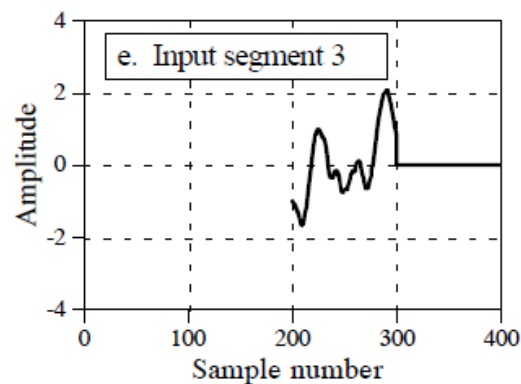
**Sinal original: 300 amostras**  
**Complementado com 100 zeros**



+



+





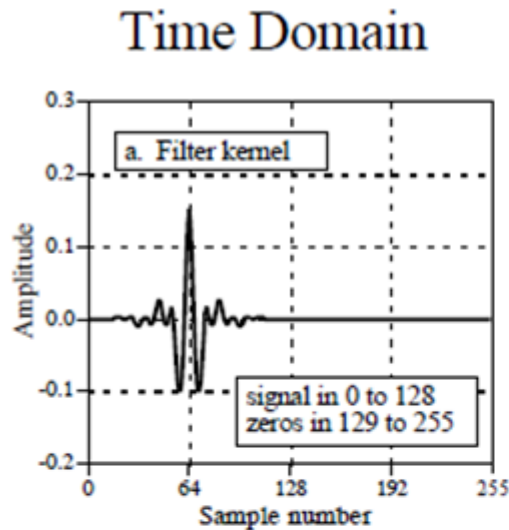
## Convolução por FFT

Considerando que:

- A FFT é rápida
- $a(t) * b(t) \iff A[f] \times B[f]$
- Método de sobreposição
- Abordagem mais rápida para kernel com  $M > 64$
- Convolução por FFT em 4 etapas
  1. Obtém a DFT (via FFT) do kernel do filtro
  2. Obtém a DFT (via FFT) do sinal de entrada
  3. Multiplica os dois sinais no domínio da frequência
  4. Obtém o sinal de saída através da DFT Inversa (via FFT Inversa - IFFT) do sinal produto

# Convolução por FFT

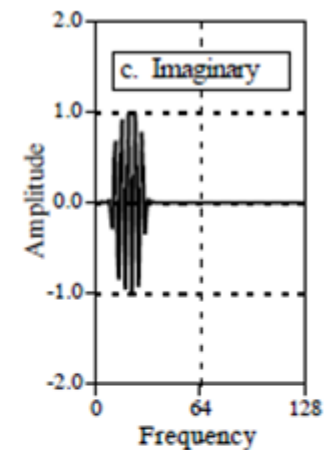
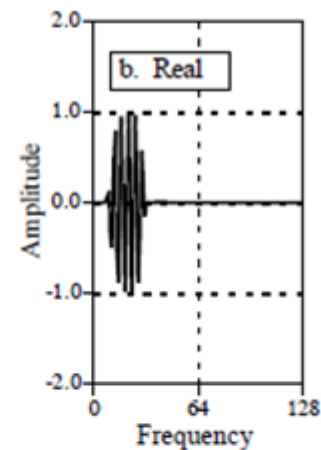
1ª Etapa



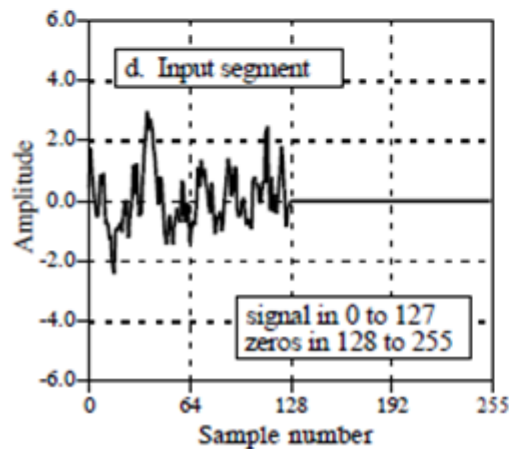
FFT



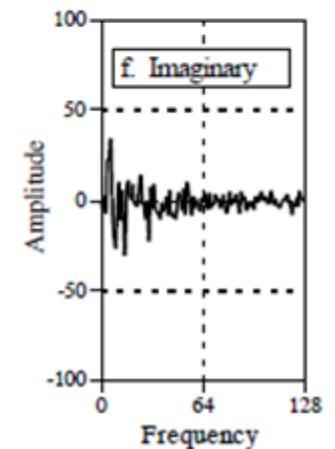
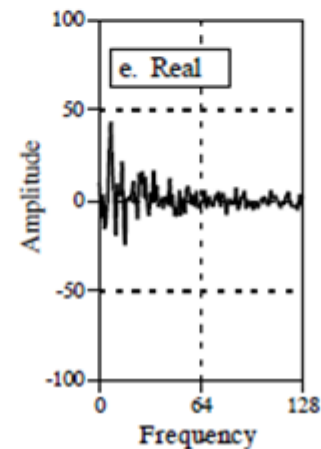
Frequency Domain



2ª Etapa

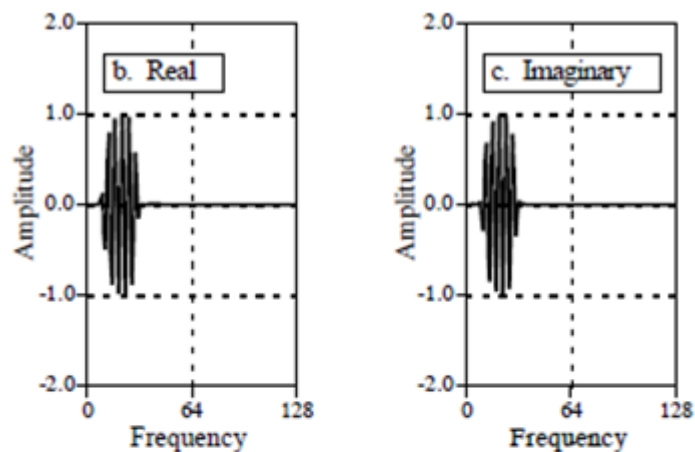


FFT

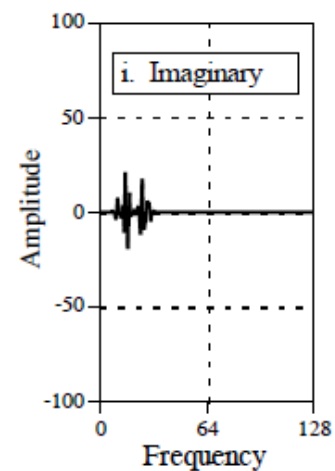
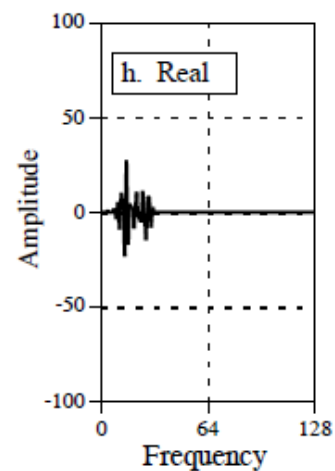
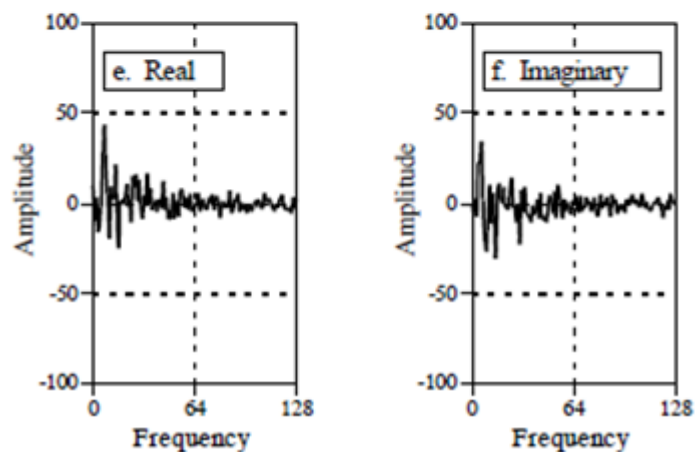


# Convolução por FFT

## Frequency Domain



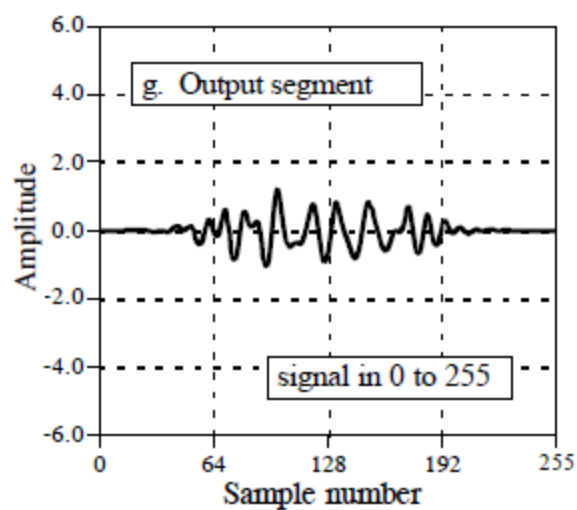
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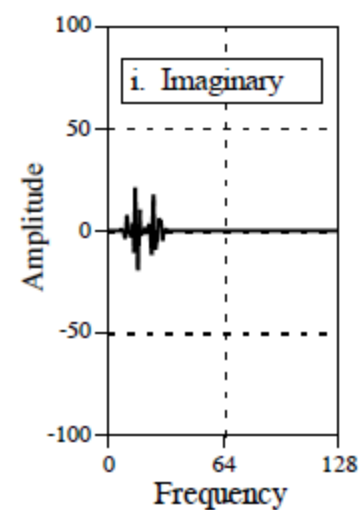
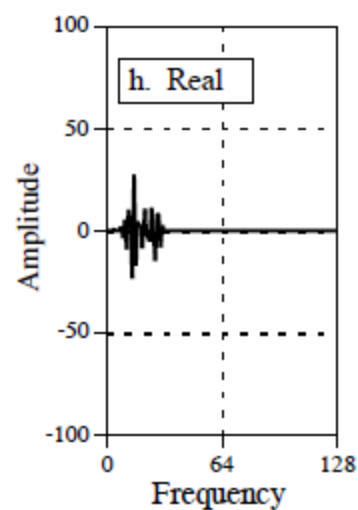
3ª Etapa

# Convolução por FFT

4ª Etapa



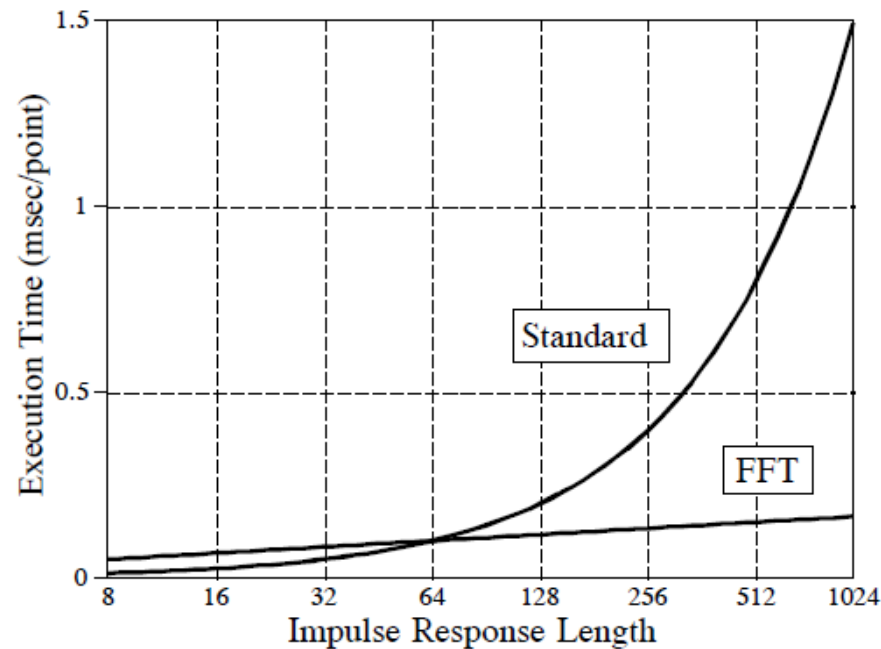
IFFT



# Convolução por FFT

## Considerações sobre desempenho:

- Filtros com kernel longo pode ser construído sem perda de performance
  - Kernel com  $M=16000$  apenas 2x mais lento que kernel com  $M=64$
- Quanto mais rápido (menos iterações), mais preciso



## Convolução por FFT

```
100 'FFT CONVOLUTION
110 'This program convolves a 10 million point signal with a 400 point filter kernel. The input
120 'signal is broken into 16000 segments, each with 625 points. 1024 point FFTs are used.
130 '
130 '
140 DIM XX[1023]
150 DIM REX[512]
160 DIM IMX[512]
170 DIM REFR[512]
180 DIM IMFR[512]
190 DIM OLAP[398]
200 '
210 FOR I% = 0 TO 398
220 OLAP[I%] = 0
230 NEXT I%
240 '
250 '
260 GOSUB XXXX
270 GOSUB XXXX
280 FOR F% = 0 TO 512
290 REFR[F%] = REX[F%]
300 IMFR[F%] = IMX[F%]
310 NEXT F%
320 '
330 '
340 FOR SEGMENT% = 0 TO 15999
350 '
360 GOSUB XXXX
370 GOSUB XXXX
380 '
390 FOR F% = 0 TO 512 'Multiply the frequency spectrum by the frequency response
400 TEMP = REX[F%]*REFR[F%] - IMX[F%]*IMFR[F%]
410 IMX[F%] = REX[F%]*IMFR[F%] + IMX[F%]*REFR[F%]
420 REX[F%] = TEMP
430 NEXT F%
440 '
450 GOSUB XXXX
460 '
470 FOR I% = 0 TO 398 'Add the last segment's overlap to this segment
480 XX[I%] = XX[I%] + OLAP[I%]
490 NEXT I%
500 '
510 FOR I% = 625 TO 1023
520 OLAP[I%-625] = XX[I%]
530 NEXT I%
540 '
550 GOSUB XXXX
560 '
570 '
580 NEXT SEGMENT%
590 '
600 GOSUB XXXX
610 END
```

'INITIALIZE THE ARRAYS  
'the time domain signal (for the FFT)  
'real part of the frequency domain (for the FFT)  
'imaginary part of the frequency domain (for the FFT)  
'real part of the filter's frequency response  
'imaginary part of the filter's frequency response  
'holds the overlapping samples from segment to segment

'zero the array holding the overlapping samples

'FIND & STORE THE FILTER'S FREQUENCY RESPONSE  
'Mythical subroutine to load the filter kernel into XX[ ]  
'Mythical FFT subroutine: XX[ ] --> REX[ ] & IMX[ ]  
'Save the frequency response in REFR[ ] & IMFR[ ]

'PROCESS EACH OF THE 16000 SEGMENTS  
'Mythical subroutine to load the next input segment into XX[ ]  
'Mythical FFT subroutine: XX[ ] --> REX[ ] & IMX[ ]

'Mythical IFFT subroutine: REX[ ] & IMX[ ] --> XX[ ]

'Save the samples that will overlap the next segment

'Mythical subroutine to output the 625 samples stored  
'in XX[0] to XX[624]

'Mythical subroutine to output all 399 samples in OLAP[ ]