

# Linear System Decomposition into FOT-Deterministic and Residual Components

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## Outline

- Fraction-of-time (FOT) probability: Almost-periodic distribution function and expectation operator
- FOT probability: Deterministic and random signals
- Spectral representation of finite average-power signals
- Linear time-variant systems
- FOT-deterministic linear systems
- Linear system decomposition into deterministic and residual components
- Conclusion
- Bibliography

## Almost-Periodic Component Extraction Operator

Let  $\mathcal{A}$  be the class of functions  $x(t)$  such that the Fourier coefficient

$$x^\alpha \triangleq \lim_{T \rightarrow \infty} \frac{1}{T} \int_{t_0 - T/2}^{t_0 + T/2} x(t) e^{-j2\pi\alpha t} dt \quad \text{exists } \forall \alpha \in \mathbb{R}$$

For every function  $x(t) \in \mathcal{A}$ , the *almost-periodic component extraction operator*  $E^{\{\alpha\}}\{\cdot\}$  extracts all the finite-strength additive sine-wave components of its argument

$$E^{\{\alpha\}}\{x(t)\} = \sum_{\alpha \in A_x} x^\alpha e^{j2\pi\alpha t}$$

$A_x =$  countable set

## Almost-Periodic Component Extraction Operator (cont'd)

### *Signal Decomposition (Gardner 1987) (Gardner Brown 1991)*

Every function  $x(t)$  of the class  $\mathcal{A}$  can be decomposed into its almost-periodic component and a residual term not containing any finite-strength additive sine-wave component

$$\begin{aligned} x(t) &= E^{\{\alpha\}}\{x(t)\} + x_r(t) \\ &= \sum_{\alpha \in A_x} x^\alpha e^{j2\pi\alpha t} + x_r(t) \end{aligned}$$

with

$$\lim_{T \rightarrow \infty} \frac{1}{T} \int_{t_0 - T/2}^{t_0 + T/2} x_r(t) e^{-j2\pi\alpha t} dt = 0 \quad \forall \alpha \in \mathbb{R}$$

## Almost-Periodic Distribution Function and Expectation Operator

*Distribution Function (Gardner 1987) (Gardner Brown 1991)*

The function

$$F_{x(t)}^{\{\alpha\}}(\xi) \triangleq E^{\{\alpha\}} \{ \mathbf{1}_{\{x(t) \leq \xi\}} \}$$

- as a function of  $\xi$ , for any  $t$ , is a valid cumulative *distribution function* except for the right-continuity property (in the discontinuity points)
- as a function of  $t$ , for any  $\xi$ , is almost periodic

*Fundamental Theorem of Temporal Expectation (Gardner 1987) (Gardner Brown 1991)*

$$E^{\{\alpha\}} \{ g(x(t)) \} = \int_{\mathbb{R}} g(\xi) dF_{x(t)}^{\{\alpha\}}(\xi)$$

The almost-periodic component extraction operator  $E^{\{\alpha\}} \{ \cdot \}$  is the *expectation operator* of the distribution function  $F_{x(t)}^{\{\alpha\}}(\xi)$  for the signal  $x(t)$ .

## Deterministic and Random Signals

$$x(t) \text{ almost-periodic function} \Leftrightarrow x(t) \equiv E^{\{\alpha\}}\{x(t)\}$$

The *almost-periodic functions* are the *deterministic* signals in the almost-periodically time-variant FOT approach. All other signals are the *random* signals.

## Spectral Representation

*Spectral Representation for Finite Average-Power signals (Wiener 1930), (Doob 1949)*

$$x(t) = \int_{\mathbb{R}} e^{j2\pi ft} dZ_x(f)$$

$Z_x(f)$  = *integrated spectrum of*  $x(t)$

The above equality means

$$x(t) = \lim_{\Delta f \rightarrow 0} \text{l.i.m.}_{B \rightarrow \infty} \int_{-B/2}^{B/2} e^{j2\pi ft} \frac{Z_x(f + \Delta f/2) - Z_x(f - \Delta f/2)}{\Delta f} df$$

$$\lim_{\Delta f \rightarrow 0} \left[ \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \left| x(t) - \lim_{B \rightarrow \infty} \int_{-B/2}^{B/2} e^{j2\pi ft} \frac{Z_x(f + \Delta f/2) - Z_x(f - \Delta f/2)}{\Delta f} df \right|^2 dt \right] = 0$$

## Spectral Representation (cont'd)

The Fourier transform  $X(f)$  of the finite-power signal  $x(t)$  can be *formally* defined by

$$dZ_x(f) = X(f) df$$

and we can *formally* write

$$x(t) = \int_{\mathbb{R}} X(f) e^{j2\pi ft} df$$

$$X(f) = \int_{\mathbb{R}} x(t) e^{-j2\pi ft} dt$$



## Linear Time-Variant Systems

### *Input/Output Relationship in the Time Domain*

$$y(t) = \int_{\mathbb{R}} h(t, u) x(u) du$$

### *Input/Output Relationship in the Frequency Domain*

$$\begin{aligned} Y(f) &\triangleq \int_{\mathbb{R}} y(t) e^{-j2\pi ft} dt \\ &= \int_{\mathbb{R}} H(f, \lambda) X(\lambda) d\lambda \end{aligned}$$

The *transmission function*  $H(f, \lambda)$  is the double Fourier transform of the *impulse-response function*  $h(t, u)$ :

$$H(f, \lambda) \triangleq \int_{\mathbb{R}^2} h(t, u) e^{-j2\pi(ft - \lambda u)} dt du$$

## FOT-Deterministic Systems

**Definition (Izzo Napolitano 2002):** An *FOT-deterministic system* is a possibly complex (and not necessarily linear) system that for every deterministic (i.e., almost periodic) input signal delivers a deterministic output signal.

$$x_\lambda(t) = e^{j2\pi\lambda t} \Rightarrow y_\lambda(t) = \sum_{\sigma \in \mathbb{S}} G_\sigma(\lambda) e^{j2\pi\varphi_\sigma(\lambda)t}$$

$\mathbb{S}$  = countable set

$\varphi_\sigma(\lambda)$  = output frequencies

$G_\sigma(\lambda)$  = Fourier coefficients

**Definition:** A *FOT-random systems* is a system that is not FOT-deterministic.

## FOT-Deterministic Linear Systems

### *FOT-Deterministic Systems*

$$x_\lambda(t) = e^{j2\pi\lambda t} \Rightarrow y_\lambda(t) = \sum_{\sigma \in \mathbb{S}} G_\sigma(\lambda) e^{j2\pi\varphi_\sigma(\lambda)t}$$

### *Homogeneity Property of Linear Systems*

$$e^{j2\pi\lambda t} X(\lambda) d\lambda \Rightarrow \sum_{\sigma \in \mathbb{S}} G_\sigma(\lambda) e^{j2\pi\varphi_\sigma(\lambda)t} X(\lambda) d\lambda$$

### *Additivity Property of Linear Systems*

$$x(t) = \int_{\mathbb{R}} X(\lambda) e^{j2\pi\lambda t} d\lambda \Rightarrow y(t) = \int_{\mathbb{R}} \sum_{\sigma \in \mathbb{S}} G_\sigma(\lambda) e^{j2\pi\varphi_\sigma(\lambda)t} X(\lambda) d\lambda$$

## FOT-Deterministic Linear Systems (cont'd)

$$\begin{aligned}
 Y(f) &= \int_{\mathbb{R}} y(t) e^{-j2\pi ft} dt \\
 &= \int_{\mathbb{R}} \sum_{\sigma \in \mathcal{S}} G_{\sigma}(\lambda) \underbrace{\int_{\mathbb{R}} e^{j2\pi[\varphi_{\sigma}(\lambda) - f]t} dt}_{\delta(f - \varphi_{\sigma}(\lambda))} X(\lambda) d\lambda \\
 &= \int_{\mathbb{R}} \underbrace{\sum_{\sigma \in \mathcal{S}} G_{\sigma}(\lambda) \delta(f - \varphi_{\sigma}(\lambda))}_{H(f, \lambda)} X(\lambda) d\lambda
 \end{aligned}$$

## FOT-Deterministic Linear Systems (cont'd)

*Transmission Function (Claasen and Mecklenbräuker 1982) (Izzo Napolitano 2002)*

$$\begin{aligned}
 H(f, \lambda) &= \sum_{\sigma \in \mathcal{S}} G_{\sigma}(\lambda) \delta(f - \varphi_{\sigma}(\lambda)) \\
 &= \sum_{\sigma \in \mathcal{S}} H_{\sigma}(f) \delta(\lambda - \psi_{\sigma}(f))
 \end{aligned}$$

$\psi_{\sigma}(\cdot)$  = inverse function of  $\varphi_{\sigma}(\cdot)$  (that can always be chosen invertible)  
 = *frequency mapping function*

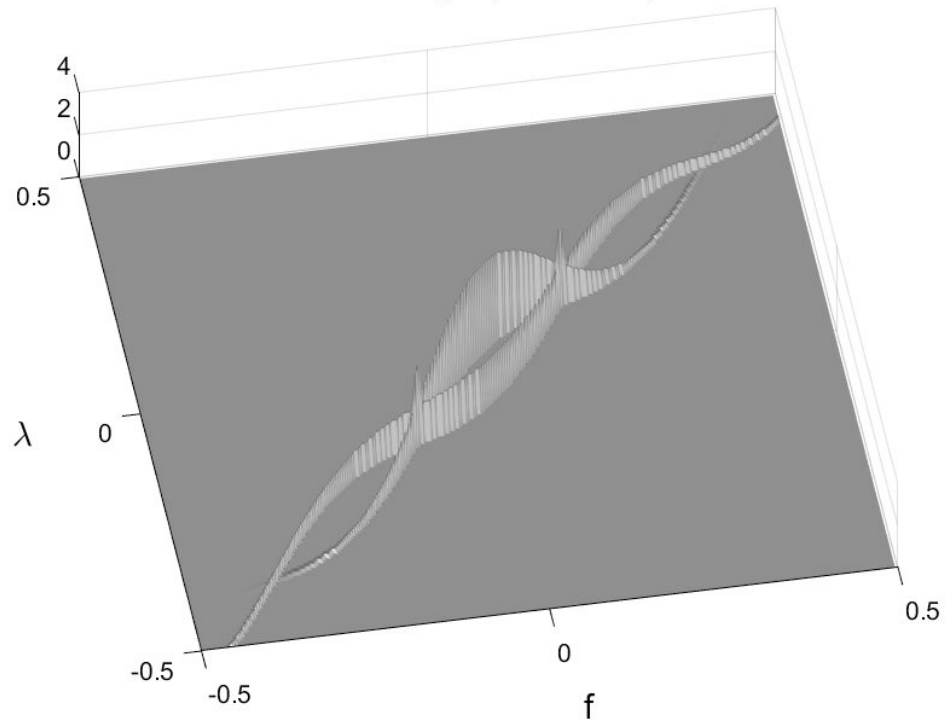
$$H_{\sigma}(f) = |\psi'_{\sigma}(f)| G_{\sigma}(\psi_{\sigma}(f))$$

$$G_{\sigma}(\lambda) = |\varphi'_{\sigma}(\lambda)| H_{\sigma}(\varphi_{\sigma}(\lambda))$$

$\psi'_{\sigma}(\cdot)$  and  $\varphi'_{\sigma}(\cdot)$  = first-order derivatives of  $\psi_{\sigma}(\cdot)$  and  $\varphi_{\sigma}(\cdot)$

## FOT-Deterministic Linear Systems (cont'd)

$$H(f, \lambda) = \sum_{\sigma} H_{\sigma}(f) \delta(\lambda - \psi_{\sigma}(f))$$



## FOT-Deterministic Linear Systems (cont'd)

*Impulse Response Function (Izzo Napolitano 2002)*

$$\begin{aligned}
 h(t, u) &\triangleq \int_{\mathbb{R}^2} H(f, \lambda) e^{j2\pi(ft - \lambda u)} \, df \, d\lambda \\
 &= \sum_{\sigma \in \mathcal{S}} \int_{\mathbb{R}} G_{\sigma}(\lambda) e^{j2\pi\varphi_{\sigma}(\lambda)t} e^{-j2\pi\lambda u} \, d\lambda \\
 &= \sum_{\sigma \in \mathcal{S}} \int_{\mathbb{R}} H_{\sigma}(f) e^{-j2\pi\psi_{\sigma}(f)u} e^{j2\pi ft} \, df
 \end{aligned}$$

## FOT-Deterministic Linear Systems (cont'd)

### *Input/Output Relationship in the Frequency Domain*

$$\begin{aligned} Y(f) &= \sum_{\sigma \in \mathcal{S}} \int_{\mathbb{R}} G_{\sigma}(\lambda) \delta(f - \varphi_{\sigma}(\lambda)) X(\lambda) d\lambda \\ &= \sum_{\sigma \in \mathcal{S}} H_{\sigma}(f) X(\psi_{\sigma}(f)) \end{aligned}$$

### *Input/Output Relationship in the Time Domain*

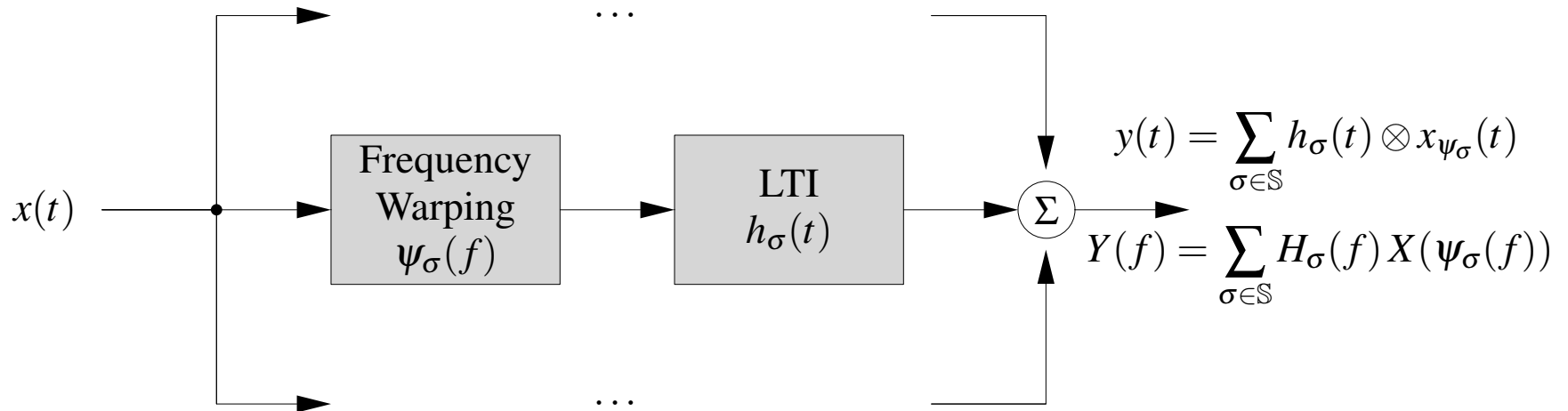
$$\begin{aligned} y(t) &= \sum_{\sigma \in \mathcal{S}} \int_{\mathbb{R}} G_{\sigma}(\lambda) X(\lambda) e^{j2\pi\varphi_{\sigma}(\lambda)t} d\lambda \\ &= \sum_{\sigma \in \mathcal{S}} h_{\sigma}(t) \otimes x_{\psi_{\sigma}}(t) \end{aligned}$$

$$x_{\psi_{\sigma}}(t) = \textit{Frequency-Warped Version of } x(t)$$

$$x_{\psi_{\sigma}}(t) \triangleq \int_{\mathbb{R}} X(\psi_{\sigma}(f)) e^{j2\pi ft} df$$



## FOT-Deterministic Linear Systems (cont'd)



$$h_\sigma(t) \xleftrightarrow{\mathcal{F}} H_\sigma(f)$$

## FOT-Deterministic Linear Systems (cont'd)

### *Examples of FOT-Deterministic LTV Systems:*

- Linear Almost-Periodically Time-Variant (LAPTIV) Systems

$$\psi_{\sigma}(f) = f + \sigma \quad \sigma \in \mathbb{S} \text{ set of frequency shifts}$$

$$y(t) = \sum_{\sigma \in \mathbb{S}} \left[ x(t) e^{j2\pi\sigma t} \right] \otimes h_{\sigma}(t)$$

- Linear Time-Invariant Systems

$$\psi_{\sigma}(f) = f \quad |\mathbb{S}| = 1$$

$$y(t) = x(t) \otimes h(t)$$

- System Performing a Time-Scale Change

$$h(t, u) = \delta(u - st) \quad \psi_{\sigma}(f) = f/s, H_{\sigma}(f) = 1 \quad |\mathbb{S}| = 1$$

$$y(t) = x(st)$$

- Decimators and interpolators are discrete-time deterministic LTV systems.

## FOT-Deterministic Linear Systems (cont'd)

### *Examples of FOT-Random LTV Systems:*

- Product Modulation

$$h(t, u) = c(t) \delta(u - t)$$

with  $c(t)$  almost-cyclostationary (not periodic or almost-periodic) or with finite energy.

$$y(t) = c(t) x(t)$$

- Time Warping

$$h(t, u) = \delta(u - \zeta(t))$$

with  $\zeta(t)$  not almost-periodic or affine.

$$y(t) = x(\zeta(t))$$

## FOT-Deterministic Linear Systems (cont'd)

### *Remarks:*

- The parallel and cascade concatenation of FOT-deterministic LTV systems is still a FOT-deterministic LTV system
- An FOT-deterministic linear system is *causal* if and only if it is LAPTV (Claasen and Mecklenbräuker 1982)
- If  $|\mathcal{S}| = 1$  (only one branch), in the *stochastic approach*, FOT-deterministic linear systems transform input wide-sense stationary stochastic processes into output wide-sense stationary processes (Franaszek 1967), (Franaszek and Liu 1967)

## Linear System Decomposition into FOT-Deterministic and Residual Components

### *Stochastic Approach*

#### *Decomposition of a Random Process into Deterministic and Residual Components*

$$X(t; \omega) = E \{X(t; \omega)\} + X_r(t; \omega) \quad t \in \mathbb{R}, \omega \in \Omega_x$$

deterministic component =  $E \{X(t; \omega)\}$  = expected value of  $X(t; \omega)$

residual component =  $X_r(t; \omega)$  with  $E \{X(t; \omega)\} = 0$

#### *Decomposition of a Random Linear System into Deterministic and Residual Components*

$$h(t, u; \omega) = E \{h(t, u; \omega)\} + h_r(t, u; \omega) \quad t, u \in \mathbb{R}, \omega \in \Omega_h$$

deterministic component =  $E \{h(t, u; \omega)\}$  = expected value of  $h(t, u; \omega)$

residual component =  $h_r(t, u; \omega)$  with  $E \{h_r(t, u; \omega)\} = 0$

## Linear System Decomposition into FOT-Deterministic and Residual Components (cont'd)

Under the assumption that the stochastic process  $\{X(t; \omega_x), t \in \mathbb{R}, \omega_x \in \Omega_x\}$  and the stochastic system  $\{h(t, u; \omega_h), t, u \in \mathbb{R}, \omega_h \in \Omega_h\}$  are *uncorrelated* ( $E\{Xh\} = E\{X\}E\{h\}$ )

$$\begin{aligned}
 E\{Y(t; \omega_x, \omega_h)\} &= E\left\{ \int_{\mathbb{R}} h(t, u; \omega_h) X(u; \omega_x) du \right\} \\
 &= E\left\{ \int_{\mathbb{R}} \left[ E\{h(t, u; \omega_h)\} + h_r(t, u; \omega_h) \right] \left[ E\{X(u; \omega_x)\} + X_r(u; \omega_x) \right] du \right\} \\
 &= \int_{\mathbb{R}} E\{h(t, u; \omega_h)\} E\{X(u; \omega_x)\} du
 \end{aligned}$$

Three terms disappear since  $E\{X_r\} = 0$ ,  $E\{h_r\} = 0$ ,  $E\{X_r h_r\} = 0$ .

## Linear System Decomposition into FOT-Deterministic and Residual Components (cont'd)

### *FOT Approach*

#### *Decomposition of a Signal into Deterministic and Residual Components*

$$x(t) = E^{\{\alpha\}} \{x(t)\} + x_r(t)$$

deterministic component =  $E^{\{\alpha\}} \{x(t)\}$  = almost-periodic component of  $x(t)$  = expected value of  $x(t)$

residual component =  $x_r(t)$  = residual term not containing any finite-strength additive sine-wave component

#### *Decomposition of a Linear System into Deterministic and Residual Components*

$$h(t, u) = h_D(t, u) + h_r(t, u)$$

deterministic component =  $h_D(t, u)$  = **FOT-deterministic system.**

It cannot be defined in terms the almost-periodic component extraction operator.

residual component =  $h_r(t, u) \triangleq h(t, u) - h_D(t, u)$  = **residual term**

## Linear System Decomposition into FOT-Deterministic and Residual Components (cont'd)

$$\begin{aligned}
 \int_{\mathbb{R}} h(t, u) x(u) \, du &= \int_{\mathbb{R}} \left[ h_D(t, u) + h_r(t, u) \right] \left[ E_u^{\{\alpha\}} \{x(u)\} + x_r(u) \right] \, du \\
 &= \int_{\mathbb{R}} h_D(t, u) E_u^{\{\alpha\}} \{x(u)\} \, du && \text{(a)} \\
 &\quad + \int_{\mathbb{R}} h_D(t, u) x_r(u) \, du && \text{(b)} \\
 &\quad + \int_{\mathbb{R}} h_r(t, u) E_u^{\{\alpha\}} \{x(u)\} \, du && \text{(c)} \\
 &\quad + \int_{\mathbb{R}} h_r(t, u) x_r(u) \, du && \text{(d)}
 \end{aligned}$$



## Linear System Decomposition into FOT-Deterministic and Residual Components (cont'd)

- (a)  $h_D(t, u)$  transforms an almost-periodic input signal into an almost-periodic output signal  
⇒ **term (a) is almost-periodic**
- (b)  $x_r(u)$  does not contain any finite-strength additive sinewave component  
⇒ passing through  $h_D(t, u)$  does not produce any output finite-strength sine wave  
⇒ **term (b) does not contain any almost-periodic component**
- (c)  $h_r(t, u)$  is the residual term with respect to  $h_D(t, u)$   
⇒ if its input is almost-periodic its output does not contain any finite-strength sine wave  
⇒ **term (c) does not contain any almost-periodic component**
- (d) If  $h_r(t, u)$  is *functionally independent* of  $x_r(u)$ , that is, there is no operator  $\mathcal{H}$  such that  $h_r(t, u) = \mathcal{H}[x_r(\cdot); t, u]$ , thus from the residual term  $x_r(u)$  no hidden higher-order almost-periodic component can be regenerated by  $h_r(t, u)$   
⇒ **term (d) does not contain any almost-periodic component**

## Linear System Decomposition into FOT-Deterministic and Residual Components (cont'd)

$$\begin{aligned}
 E_t^{\{\alpha\}} \{y(t)\} &= E_t^{\{\alpha\}} \left\{ \int_{\mathbb{R}} h(t, u) x(u) du \right\} \\
 &= E_t^{\{\alpha\}} \left\{ \int_{\mathbb{R}} \left[ h_D(t, u) + h_r(t, u) \right] \left[ E_u^{\{\alpha\}} \{x(u)\} + x_r(u) \right] du \right\} \\
 &= \int_{\mathbb{R}} h_D(t, u) E_u^{\{\alpha\}} \{x(u)\} du
 \end{aligned}$$

$h_D(t, u)$  is the kernel of the linear operator that transforms the almost-periodic component of the input signal into the almost-periodic component of the output signal.

## Linear System Decomposition into FOT-Deterministic and Residual Components (cont'd)

	<i>FOT deterministic</i>	<i>FOT random</i>	<i>residual term</i>
<i>signals</i>	almost-periodic (AP) functions	not FOT deterministic	not containing any finite-strength additive sine-wave component
<i>systems</i>	transform AP functions into AP functions	not FOT deterministic	transform AP functions into signals not containing finite-strength additive sine-wave components

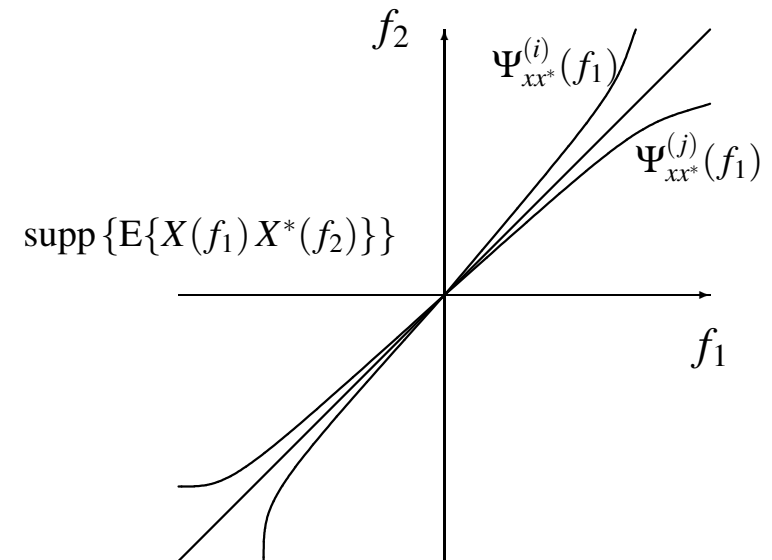
## FOT-Deterministic LTV Systems and Spectrally Correlated Processes

### *Spectrally Correlated (SC) Processes*

*Loève bifrequency spectrum*

$$E \{X(f_1) X^*(f_2)\} = \sum_k S_{xx^*}^{(k)}(f_1) \delta(f_2 - \Psi_{xx^*}^{(k)}(f_1))$$

Spectral masses are concentrated on a countable set of support curves



## **FOT-Deterministic LTV Systems and Spectrally Correlated Processes (cont'd)**

- Spectrally correlated processes are closed under FOT-deterministic LTV transformations.
- The Wiener filter with desired and input signal that are singularly and jointly spectrally correlated is a FOT-deterministic LTV system.

## Conclusion

- In the FOT approach signals can be decomposed into the sum of an almost-periodic component and a residual term non containing any finite-strength additive sine-wave component.
- In the FOT approach linear systems can be decomposed into the parallel of a FOT deterministic component and a residual term.
- The FOT deterministic component transforms input almost-periodic functions into output almost-periodic functions.
- The residual term transforms input sine waves into output signals not containing any finite-strength additive sine-wave component.

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