

# Sparse Recovery with Finite Rate of Innovation Sampling

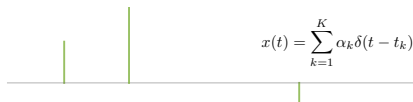
Hanjie Pan, Frederike Dümbgen

École Polytechnique Fédérale de Lausanne

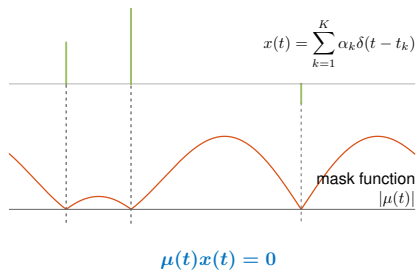
MFSP Application Session  
December 17, 2018



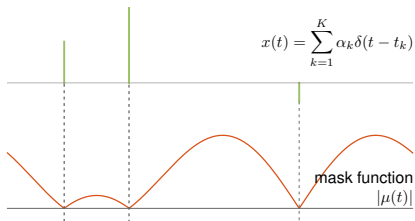
# Sensing Sparsity through Annihilation



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$$\mu(t)x(t) = 0$$

- Dirac locations: **zero-crossings** of a mask function, e.g.,

$$\mu(t) = \underbrace{\prod_{k=1}^K (1 - e^{-j \frac{2\pi}{\tau} t_k} e^{j \frac{2\pi}{\tau} t})}_{\text{poly}\left(e^{j \frac{2\pi}{\tau} t}\right)}$$

# A Generalized FRI Sampling Framework

**Goal:** reconstruct an FRI signal that is *consistent* with the given measurements.

[1] H. Pan, T. Blu, and M. Vetterli. “Towards Generalized FRI Sampling with an Application to Source Resolution in Radioastronomy”. In: *IEEE Transactions on Signal Processing* 65.4 (2017), pp. 821–835.

# A Generalized FRI Sampling Framework

**Goal:** reconstruct an FRI signal that is *consistent* with the given measurements.

- 1 What is an FRI signal?
- 2 What is a consistent reconstruction?

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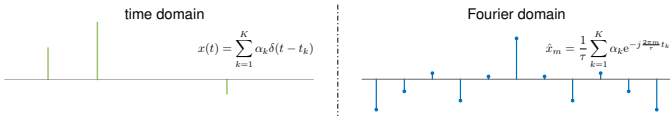

$$x(t) = \sum_{k=1}^K \alpha_k \delta(t - t_k)$$

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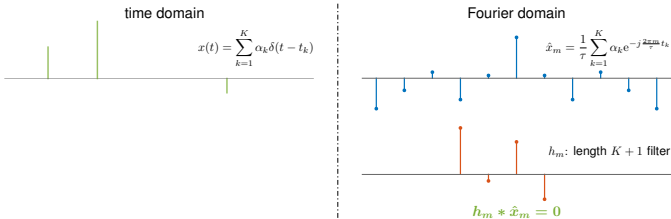
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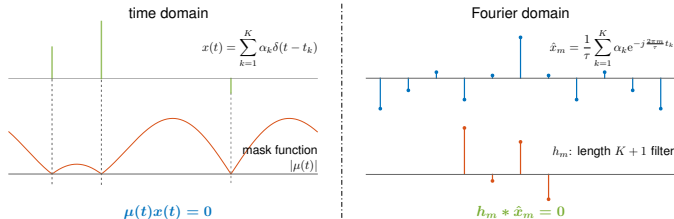


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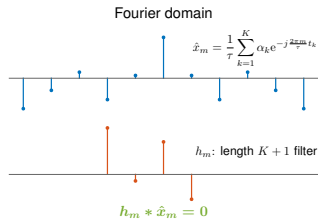
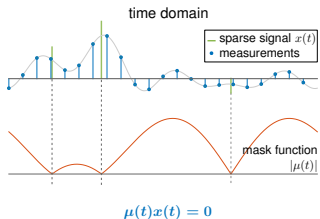


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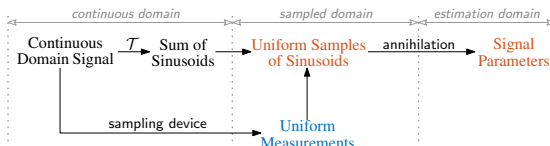


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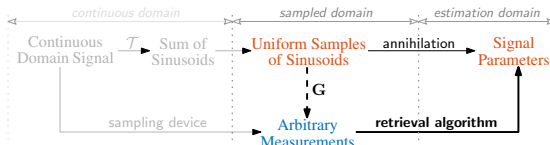


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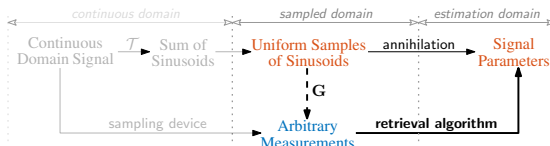


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# A Generalized FRI Sampling Framework

**Goal:** reconstruct an FRI signal that is *consistent* with the given measurements.

## 1 What is an FRI signal?



## 2 What is a consistent reconstruction?

$$\|\mathbf{y} - \mathbf{y}'\|_2 \leq \varepsilon$$

- $\mathbf{y}$ : the given measurements
- $\mathbf{y}'$ : the re-synthesized measurements
- $\varepsilon$ : noise level

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## Problem 1: Constrained Minimization

$$\begin{aligned} \min_{\mathbf{h} \in \mathcal{H}, \mathbf{b}} \quad & \|\mathbf{y} - \mathbf{G}\mathbf{b}\|_2^2 \\ \text{subject to} \quad & \mathbf{b} * \mathbf{h} = \mathbf{0} \end{aligned}$$

- $\mathbf{y}$ : the given set of measurements
- $\mathbf{b}$ : the unknown uniform samples of sinusoids
- $\mathbf{h}$ : the annihilating filter coefficients
- $\mathbf{G}$ : linear transformation from  $\mathbf{b}$  to  $\mathbf{y}$

## Problem 2: Constrained Approximation with Noise Level ( $\varepsilon^2$ )

find  $\mathbf{h} \in \mathcal{H}, \mathbf{b}$

subject to  $\mathbf{b} * \mathbf{h} = \mathbf{0}$

$$\|\mathbf{y} - \mathbf{G}\mathbf{b}\|_2^2 \leq \varepsilon^2$$

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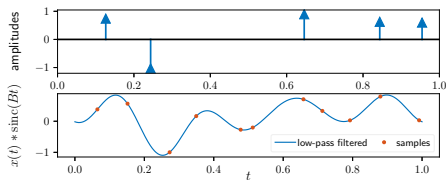
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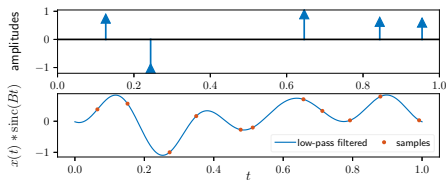
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# Example: Dirac Estimation (Time-domain)



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- 1 Uniform sinusoidal samples **b** — FS coefficients

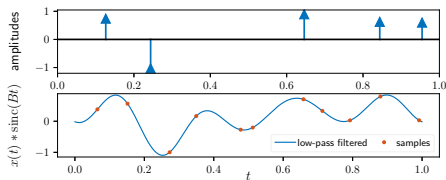
$$\hat{x}_m = \frac{1}{\tau} \sum_{k=1}^K \alpha_k e^{-j \frac{2\pi t_k m}{\tau}}$$

- 2 Relation with given measurements **G** — inverse DFT

$$y_\ell = \frac{1}{B} \sum_{m \in \mathcal{M}} \hat{x}_m e^{j \frac{2\pi m}{\tau} t'_\ell}$$

$t'_\ell$ : sampling locations,  $\mathcal{M} = \{m \mid |m| \leq \lfloor \tau B/2 \rfloor\}$ .

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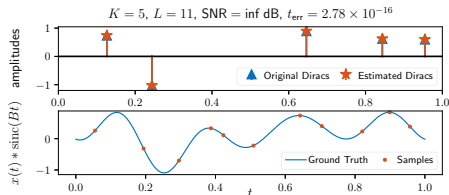
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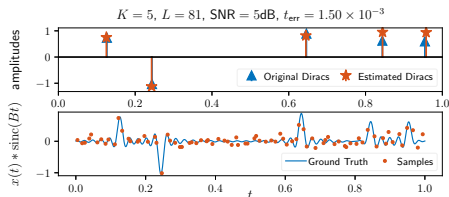
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# Example: Dirac Estimation (Time-domain)



(a) noiseless reconstruction



(b) noisy reconstruction

Figure: Dirac reconstruction from non-uniform time-domain samples.