

Time-Varying Linear Systems: Identification and Transmission through Unidentified Channels

Alihan Kaplan[†], Dae Gwan Lee*, Götz Pfander*, Volker Pohl[†]

*Lehrstuhl für Wissenschaftliches Rechnen, Katholische Universität Eichstätt-Ingolstadt

[†]Lehrstuhl für Theoretische Informationstechnik, Technische Universität, München



Outline

1. Introduction: Linear Time-Varying Systems
2. Identification under Side Constraints
3. Signal Transmission over Unidentified Channels
4. Example: Message Transmission with unknown Channel Support

Time-Varying Linear Systems

Time-varying Linear SISO Systems

▷ Single-Input Single-Output Systems (SISO):

Linear time-varying SISO channels are described by operators of $H : L^2(\mathbb{R}) \rightarrow L^2(\mathbb{R})$ the form

$$(Hf)(t) = \iint_{\mathbb{R} \times \mathbb{R}} \eta_H(\tau, \nu) \cdot f(t - \tau) e^{i2\pi\nu t} d\nu d\tau = \iint_{\mathbb{R} \times \mathbb{R}} \eta_H(\tau, \nu) (M_\nu T_\tau f)(t) d\nu d\tau$$

with

- Spreading function: $\eta_H \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{C}$
- Translation (time-shift) operator: $(T_\tau f)(t) = f(t - \tau)$
- Modulation (frequency shift): $M_\nu(f)(t) = f(t) e^{i2\pi\nu t}$

Time-varying Linear MIMO Systems

▷ **Multiple-Input Multiple-Output Systems (MIMO):**

Channels with N -inputs and M -outputs are characterized by operators $\mathbf{H} : (L^2(\mathbb{R}))^N \rightarrow (L^2(\mathbb{R}))^M$ of the form

$$\mathbf{H} \begin{pmatrix} f_1 \\ \vdots \\ f_N \end{pmatrix} = \begin{pmatrix} H_{1,1} & \cdots & H_{1,N} \\ \vdots & & \vdots \\ H_{M,1} & \cdots & H_{M,N} \end{pmatrix} \begin{pmatrix} f_1 \\ \vdots \\ f_N \end{pmatrix} = \begin{pmatrix} \sum_{n=1}^N H_{1,n} f_n \\ \vdots \\ \sum_{n=1}^N H_{M,n} f_n \end{pmatrix}.$$

Each subchannel $H_{m,n}$ is a TVL SISO system.

Finite-dimensional TVL Channels

- ▷ The identification problem of TVL systems $\mathbf{H} : (L^2(\mathbb{R}))^N \rightarrow (L^2(\mathbb{R}))^M$ can be reduced to a finite-dimensional problem $\mathbf{H} : (\mathbb{C}^L)^N \rightarrow (\mathbb{C}^L)^M$

$$\mathbf{H} \begin{pmatrix} \mathbf{x}_1 \\ \vdots \\ \mathbf{x}_N \end{pmatrix} = \begin{pmatrix} \mathbf{H}_{1,1} & \cdots & \mathbf{H}_{1,N} \\ \vdots & & \vdots \\ \mathbf{H}_{M,1} & \cdots & \mathbf{H}_{M,N} \end{pmatrix} \begin{pmatrix} \mathbf{x}_1 \\ \vdots \\ \mathbf{x}_N \end{pmatrix} = \begin{pmatrix} \sum_{n=1}^N \mathbf{H}_{1,n} \mathbf{x}_n \\ \vdots \\ \sum_{n=1}^N \mathbf{H}_{M,n} \mathbf{x}_n \end{pmatrix}.$$

wherein each sub-system $\mathbf{H}_{n,m} : \mathbb{C}^L \rightarrow \mathbb{C}^L$ has the form

$$\mathbf{H}_{m,n} \mathbf{x} = \sum_{\ell=0}^{L-1} \sum_{k=0}^{L-1} \eta_{m,n}(k, \ell) \mathbf{M}^\ell \mathbf{T}^k \mathbf{x} = \mathbf{G}(\mathbf{x}) \boldsymbol{\eta}$$

with **spreading coefficients** $\{\eta_{m,n}(k, \ell)\}_{k, \ell=0}^{L-1}$, and with the **translation operator** $\mathbf{T} : \mathbb{C}^L \rightarrow \mathbb{C}^L$ and the **modulation operator** $\mathbf{M} : \mathbb{C}^L \rightarrow \mathbb{C}^L$ given by

$$\mathbf{T} : \begin{pmatrix} x_0 \\ x_1 \\ \vdots \\ x_{L-1} \end{pmatrix} \mapsto \begin{pmatrix} x_{L-1} \\ x_0 \\ \vdots \\ x_{L-2} \end{pmatrix} \quad \text{and} \quad \mathbf{M} : \begin{pmatrix} x_0 \\ x_1 \\ \vdots \\ x_{L-1} \end{pmatrix} \mapsto \begin{pmatrix} x_0 \\ x_1 e^{i \frac{2\pi}{L} \cdot 1} \\ \vdots \\ x_{L-1} e^{i \frac{2\pi}{L} \cdot (L-1)} \end{pmatrix}$$

Operator Paley–Wiener spaces

- ▷ Set of all linear operators $\mathbb{C}^L \rightarrow \mathbb{C}^L$ is spanned by time-frequency shifts $\mathbf{M}^\ell \mathbf{T}^k$

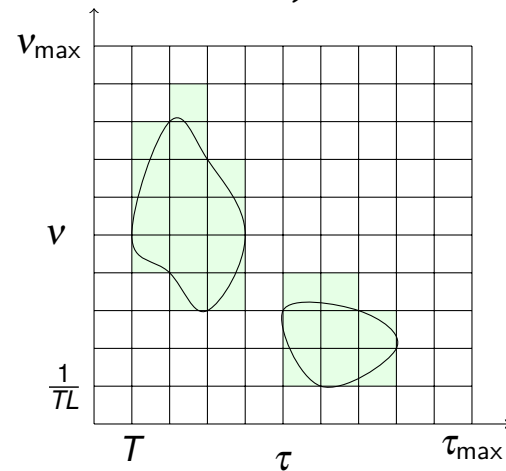
$$\mathcal{L}(\mathbb{C}^L) = \left\{ \mathbf{H} = \sum_{k=0}^{L-1} \sum_{\ell=0}^{L-1} \eta(k, \ell) \mathbf{M}^\ell \mathbf{T}^k : \eta(k, \ell) \in \mathbb{C} \text{ for all } (k, \ell) \in \mathbb{Z}_L \times \mathbb{Z}_L \right\}$$

- ▷ **SISO Operator Paley–Wiener space:** For $\Lambda \subset \mathbb{Z}_L \times \mathbb{Z}_L$

$$OPW(\Lambda) = \text{span} \left\{ \mathbf{M}^\ell \mathbf{T}^k : (k, \ell) \in \Lambda \right\}$$

- ▷ **MIMO Operator Paley–Wiener space:** For $\mathbf{\Lambda} = \{\Lambda_{m,n}\}_{m,n=1}^{M,N}$ with $\Lambda_{m,n} \in \mathbb{Z}_L \times \mathbb{Z}_L$

$$OPW(\mathbf{\Lambda}) = \{ \mathbf{H} : \mathbf{H}_{m,n} \in OPW(\Lambda_{m,n}) \}$$



Identification under linear side constraints

Identification – SISO

Definition (Identifiable)

The space $OPW(\Lambda)$ is identifiable if and only if there exists an identifier $\mathbf{c} \in \mathbb{C}^L$ such that for each $\mathbf{H} \in OPW(\Lambda)$ the equation

$$\mathbf{y} = \mathbf{H}\mathbf{c} = \sum_{(k,\ell) \in \Lambda} \eta(k,\ell) \mathbf{M}^\ell \mathbf{T}^k \mathbf{c} = \mathbf{G}(\mathbf{c})\boldsymbol{\eta}$$

is uniquely solvable for $\boldsymbol{\eta} \in \mathbb{C}^\Lambda$.

Remark

$\mathbf{G}(\mathbf{c})$ is Gabor matrix of size $L \times L^2$

$$\mathbf{G}(\mathbf{c}) = [\mathbf{M}^0 \mathbf{T}^0 \mathbf{c}, \mathbf{M}^0 \mathbf{T}^1 \mathbf{c}, \dots, \mathbf{M}^{L-1} \mathbf{T}^{L-1} \mathbf{c}]$$

Theorem (Identification of SISO Channels)

The space $OPW(\Lambda)$ is identifiable if and only if $|\Lambda| \leq L$.

Identification – MIMO

Definition (Identifiable MIMO)

The space $OPW(\Lambda)$ is identifiable if and only if there exist vectors $\mathbf{c} = (\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_N) \in (\mathbb{C}^L)^N$ such that for each $\mathbf{H} \in OPW(\Lambda)$ the map $\mathbf{H} \mapsto \mathbf{y} = \mathbf{H}\mathbf{c}$ is injective.

Theorem (Identification of MIMO Channels)

The space $OPW(\Lambda)$ is identifiable if and only if

$$\sum_{n=1}^N |\Lambda_{m,n}| \leq L \quad \text{for every } m = 1, \dots, M.$$

Assumptions

- Subchannels $\mathbf{H}_{m,n}$ and their time-frequency components $\eta_{m,n}(k, \ell)$ are independent
- Information about one channel does not help to identify another.

Linear Constraints Between Spreading Coefficients

- ▷ Channels \mathbf{H} in $OPW(\Lambda) \subset \mathcal{L}(\mathbb{C}^L)$:

$$\mathbf{y} = \mathbf{H}\mathbf{c} = \sum_{(k,\ell) \in \Lambda} \eta(k,\ell) \mathbf{M}^{\ell} \mathbf{T}^k \mathbf{c} = \mathbf{G}(\mathbf{c})\boldsymbol{\eta}$$

- ▷ $OPW(\Lambda)$ is identifiable if $|\Lambda| \leq L$.
- ▷ Assume linear relations between the spreading coefficients are known

$$\sum_{k,\ell} \alpha_{k,\ell} \eta(k,\ell) = \beta \quad \text{for some } \alpha_{k,\ell}, \beta \in \mathbb{C}$$

Intuition/Question

- Let $\mathbf{A}\boldsymbol{\eta} = \mathbf{b}$ be a given set of $M \geq 1$ linear independent side constraints.
- Let $OPW_{\mathbf{A},\mathbf{b}}(\Lambda)$ be the set of all $\mathbf{H} \in OPW(\Lambda)$ which satisfy these side constraints.
- $OPW_{\mathbf{A},\mathbf{b}}(\Lambda)$ is identifiable if and only if $|\Lambda| \leq L + M$?

Linear Relations between Time-Frequency Components

- ▷ In the SISO case, linear relations between the spreading coefficients of the channel are expressed by

$$\mathbf{b} = \mathbf{A}\boldsymbol{\eta}$$

- ▷ Including the equation for channel identification yields

$$\begin{bmatrix} \mathbf{y} \\ \mathbf{b} \end{bmatrix} = \begin{bmatrix} \mathbf{G}(\mathbf{c}) \\ \mathbf{A} \end{bmatrix} \boldsymbol{\eta}$$

→ More equations for the same number of unknowns should help to identify the channel.

Theorem (One linear side constraint)

Let $\mathbf{H} \in OPW(\Lambda)$ with $\Lambda \subset \mathbb{Z}_L \times \mathbb{Z}_L$ with $|\Lambda| = L + 1$ and let $\mathbf{a} \in \mathbb{C}^{L+1}$, $\mathbf{a} \neq \mathbf{0}$. There there exists a $\mathbf{c} \in \mathbb{C}^L$ so that

$$\begin{bmatrix} \mathbf{G}(\mathbf{c})|_{\Lambda} \\ \mathbf{a}^* \end{bmatrix}$$

is invertible. Thus the channel coefficients $\boldsymbol{\eta}$ are identifiable.

Moreover, the set of all such identifiers \mathbf{c} constitute a dense open subset of \mathbb{C}^L .

More Side Constrains

Lemma (No general solution for more than one side constraint)

Let $\mathbf{H} \in OPW(\Lambda)$ with $\Lambda \subset \mathbb{Z}_L \times \mathbb{Z}_L$ with $|\Lambda| > L + 1$. There exist matrices \mathbf{A} of size $(|\Lambda| - L) \times |\Lambda|$ with $\mathbf{A} \neq 0$ such that there is no $\mathbf{c} \in \mathbb{C}^L$ such that the matrix

$$\begin{bmatrix} \mathbf{G}(\mathbf{c})|_{\Lambda} \\ \mathbf{A} \end{bmatrix}$$

has full rank.

Theorem (Sufficient condition for identifiability)

Let $\mathbf{H} \in OPW(\Lambda)$ with $\Lambda \in \mathbb{Z}_L \times \mathbb{Z}_L$ of size $|\Lambda| = R > L$. Assume that there exists a subset $\tilde{\Lambda} \subset \Lambda$ of size L so that

- (i) $\tau_j(\Lambda) = \tau_j(\tilde{\Lambda})$ whenever $\tau_j(\Lambda) \neq 0$.
- (ii) $\text{ind}(\tau') \neq \text{ind}(\tau(\Lambda))$ for every L -tubel $\tau' \preceq \tau(\tilde{\Lambda})$ of size L different from $\tau(\Lambda)$.

Given any full spark matrix \mathbf{A} of size $(R - L) \times R$, then there exists an identifier $\mathbf{c} \in \mathbb{C}^L$ so that the $R \times R$ matrix

$$\begin{bmatrix} \mathbf{G}(\mathbf{c})|_{\Lambda} \\ \mathbf{A} \end{bmatrix}$$

is invertible. Moreover, the set of all such identifiers \mathbf{c} constitute a dense open subset of \mathbb{C}^L .

Signal Transmission over Unidentified Channels

Motivation

- ▷ Two step procedure for data transmission over frequency-selective channels
 1. Estimate (identify) the channel $\mathbf{H} \in \mathcal{H} \subset \mathcal{L}(\mathbb{C}^L)$.
 2. Transmit data \mathbf{x} from a certain data set $\mathcal{X} \subset \mathbb{C}^L$. Data recovery at the receiver using estimated channel.
- ▷ In time-varying channels, this procedure has to be repeated regularly, to update channel state information.
- ▷ In rapidly changing channels, two step procedure becomes more and more inefficient.

Transmission through unidentified channel

- ▷ Combine channel identification and signal recovery.
- ▷ Transmission scheme $\mathbf{y} = \mathbf{H}(\mathbf{x} + \mathbf{c})$ with
 - $\mathbf{H} \in \mathcal{H} \subset \mathcal{L}(\mathbb{C}^L)$ is a **unknown channel** from a known subset \mathcal{H} .
 - $\mathbf{x} \in \mathcal{X}$ is the **data signal** from a certain data set $\mathcal{X} \subset \mathbb{C}^L$.
 - $\mathbf{c} \in \mathbb{C}^L$ is a **pilot signal** (designed based on the knowledge of \mathcal{H} and \mathcal{X}).

Problem

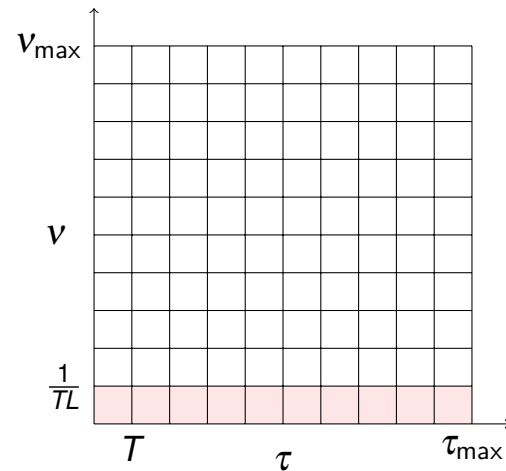
Find (necessary and/or sufficient) conditions on \mathcal{H} and \mathcal{X} such that there exists a pilot $\mathbf{c} \in \mathbb{C}^L$ such that every $\mathbf{x} \in \mathcal{X}$ can uniquely be recovered from $\mathbf{y} = \mathbf{H}(\mathbf{x} + \mathbf{c})$ for any unknown channel $\mathbf{H} \in \mathcal{H}$.

Relation to Blind Deconvolution

- ▷ $\mathcal{H} = OPW(\Lambda)$ with $\Lambda \{(0,0), (1,0), \dots, (L,0)\} \subset \mathbb{Z}_L \times \mathbb{Z}_L$

$$\mathbf{y} = \mathbf{H}\mathbf{c} = \sum_{k=0}^{L-1} \eta(k,0) \mathbf{T}^k \mathbf{x}$$

- ▷ Recover $\mathbf{x} \in \mathcal{X} \in \mathbb{C}^L$ and $\boldsymbol{\eta}$ from \mathbf{y} (without knowing η).
- ▷ \mathbf{x} and $\boldsymbol{\eta}$ assumed to be sparse.



Conditions for Data Recovery and/or Channel Identification

Natural Conditions

- (I) **Identifiability of \mathcal{H}** : The map $\mathbf{H} \mapsto \mathbf{H}\mathbf{c}$ is injective on \mathcal{H} .
- (R) **Recovery condition for known channel**: Every $\mathbf{H} \in \mathcal{H}$ is injective on \mathcal{X} .

Subsets of \mathbb{C}^L

- $\mathcal{H}\mathbf{c} = \{\mathbf{H}\mathbf{c} : \mathbf{H} \in \mathcal{H}\}$: All possible output vectors for the pilot \mathbf{c} .
- $\mathcal{H}\mathcal{X} = \{\mathbf{H}\mathbf{x} : \mathbf{H} \in \mathcal{H}, \mathbf{x} \in \mathcal{X}\}$: Possible output of arbitrary data vector in \mathcal{X} .

Further Conditions

- (i) $\text{span}\{\mathcal{H}\mathbf{c}\} \cap \text{span}\{\mathcal{H}\mathcal{X}\} = \{0\}$: Isolate $\mathbf{H}\mathbf{c}$ and $\mathbf{H}\mathbf{x}$ from the channel output $\mathbf{y} = \mathbf{H}(\mathbf{c} + \mathbf{x})$.
- (ii) $\mathbf{H}(\mathcal{X} + \mathbf{c}) \cap \mathbf{H}'(\mathcal{X} + \mathbf{c})$ for every $\mathbf{H} \neq \mathbf{H}'$ in \mathcal{H} : Identify \mathbf{H} from $\mathbf{y} = \mathbf{H}(\mathbf{c} + \mathbf{x})$ with unknown $\mathbf{x} \in \mathcal{X}$.
- (iii) $\mathbf{H}(\mathbf{x} + \mathbf{c}) = \mathbf{H}'(\mathbf{x}' + \mathbf{c})$ implies $\mathbf{x} = \mathbf{x}'$: Guarantees exact recovery of $\mathbf{x} \in \mathcal{X}$ but not identification \mathbf{H}

$$(i) \xrightarrow{(I)} (ii) \xrightarrow{(R)} (iii)$$

Degrees of Freedom

- ▷ $\mathcal{H} = OPW(\Lambda) \subset \mathcal{L}(\mathbb{C}^L)$ is a linear subspace of dimension $|\Lambda|$.
- ▷ Assume $\mathcal{X} \subset \mathbb{C}^L$ is a linear subspace of dimension K .
- ▷ Counting degrees of freedom, we must have

$$|\Lambda| + K \leq L \quad \text{and} \quad |\Lambda| \leq L. \quad (1)$$

as a necessary condition for **exact recovery of $\mathbf{H} \in \mathcal{H}$ and $\mathbf{x} \in \mathcal{X}$** .

- ▷ Without identifying \mathbf{H} , we may get $K > L - |\Lambda| \implies$ **how?**
- ▷ When we get equality in (1)?

Example: 1-Dimensional Signal Space

- ▷ For a given (and known) $\Lambda \subset \mathbb{Z}_L \times \mathbb{Z}_L$, we consider the operator Paley–Wiener space

$$\mathcal{H} = OPW(\Lambda) = \left\{ \mathbf{H} = \sum_{k, \ell \in \Lambda} \eta(k, \ell) \mathbf{M}^\ell \mathbf{T}^k : \eta(k, \ell) \in \mathbb{C} \right\} \subset \mathcal{L}(\mathbb{C}^L)$$

- ▷ Assume a 1-dimensional signal space $\mathcal{X} = \text{span}\{\mathbf{v}\} \subset \mathbb{C}^L$ for some $\mathbf{v} \in \mathbb{C}^L$.
- ▷ For $x = u\mathbf{v} \in \mathcal{X}$, with $u \in \mathbb{C}$, the received signal is

$$\mathbf{y} = \mathbf{H}(\mathbf{x} + \mathbf{c}) = u\mathbf{H}(\mathbf{v}) + \mathbf{H}(\mathbf{c}) = \mathbf{G}(\mathbf{c})|_\Lambda \boldsymbol{\eta} + u\mathbf{G}(\mathbf{v})|_\Lambda \boldsymbol{\eta}$$

- ▷ Separation of $\mathbf{H}\mathbf{x}$ and $\mathbf{H}\mathbf{c}$ from $\mathbf{y} \implies \text{span}\{\mathcal{H}\mathbf{c}\} \cap \text{span}\{\mathcal{H}\mathbf{v}\} = \{0\} \implies \text{rang}[\mathbf{G}(\mathbf{c})|_\Lambda] \perp \text{rang}[\mathbf{G}(\mathbf{v})|_\Lambda]$.

For $\text{rang}[\mathbf{G}(\mathbf{v})|_\Lambda] = \text{span}\{\mathbf{a}\}$ is a 1-dimensional subspace, we have the following result.

Theorem

Let $\Lambda \subset \mathbb{Z}_L \times \mathbb{Z}_L$ with $1 \leq |\Lambda| \leq L - 1$ and let $\mathbf{a} \in \mathbb{C}^L \setminus \{0\}$ be arbitrary.

There exists a vector $\mathbf{c} \in \mathbb{C}^L$ such that the $L \times |\Lambda| + 1$ matrix $[\mathbf{G}(\mathbf{c})|_\Lambda, \mathbf{a}]$ has full rank.

General Construction

- ▷ The set of all time-frequency shifts $\{\mathbf{M}^\ell \mathbf{T}^k\}_{\ell,k=0}^{L-1}$ can be separated into $L + 1$ commutative subgroups

$$\mathcal{G}_s = \{\mathbf{M}^{2rk} \mathbf{T}^k : k = 0, 1, \dots, L-1\}, \quad s = 0, 1, \dots, L-1$$

$$\mathcal{G}_L = \{\mathbf{M}^k : k = 0, 1, \dots, L-1\}$$

- ▷ Each commutative subgroup \mathcal{G}_s possesses a set of common eigenvectors (i.e. the *chirp sequences*)

$$\mathbf{e}_s(\ell) : \ell = 0, 1, \dots, L-1$$

which forms an orthogonal basis for \mathbb{C}^L .

Theorem

Let $\Lambda \subset \mathbb{Z}_L \times \mathbb{Z}_L$ such that there exists an $s \in \{0, 1, \dots, L\}$ so that $\Lambda \subset \mathcal{G}_s$.

There exists a subspace $\mathcal{X} = \text{span}\{\mathbf{e}_s(1), \dots, \mathbf{e}_s(K)\} \subset \mathbb{C}^L$ of dimension $K = L - |\Lambda|$ and a $\mathbf{c} \in \mathbb{C}^L$ such that

$$\text{span}\{\mathcal{H} \mathbf{c}\} \cap \text{span}\{\mathcal{H} \mathcal{X}\} = \{\mathbf{0}\}$$

and such that Conditions (I) and (R) are satisfied.

Maximum Size of Data Subspace

- ▷ Given a subspace \mathcal{X} of dimension K , it is desirable that the dimension of $\text{span}\{\mathcal{H}\mathcal{X}\}$ is again K .
- ▷ For a one-dimensional $\mathcal{X} = \text{span}\{\mathbf{v}\}$, we have $\dim(\text{span}\{\mathcal{H}\mathcal{X}\}) = \text{rank}(\mathbf{G}(\mathbf{v})|_{\Lambda})$.

Question: Given a support set $\Lambda \subset \mathbb{Z}_L \times \mathbb{Z}_L$ of $\mathcal{H} = \text{OPW}(\Lambda)$ with $|\Lambda| \leq L$.
 What is the minimum rank of $\mathbf{G}(\mathbf{v})|_{\Lambda}$ for \mathbf{v} varying in \mathbb{C}^L ?

Theorem

Let $L \geq 3$ be an odd integer and $\Lambda \subset \mathbb{Z}_L \times \mathbb{Z}_L$ with $|\Lambda| \leq L$. Then

$$\min_{\mathbf{v} \in \mathbb{C}^L \setminus \{0\}} \text{rank}(\mathbf{G}(\mathbf{v})|_{\Lambda}) \leq N(\Lambda)$$

with

$$N(\Lambda) = 1 + \min_{s \in \{0, 1, \dots, L\}} \min \left\{ |I| : I \subset \mathbb{Z}_L \times \mathbb{Z}_L \text{ with } \Lambda \subset I + \mathcal{G}_s \right\},$$

and where $I + \mathcal{G}_s = \{x + y : x \in I, y \in \mathcal{G}_s\}$.

Example with Unknown Channel Support

Channel Model

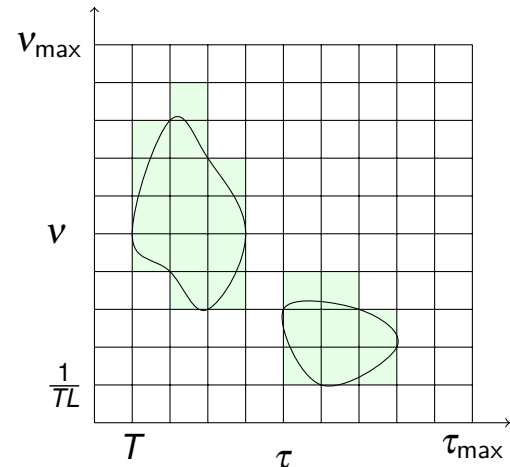
- ▷ Time continuous Time-Varying-Linear SISO Channel $H : L^2(\mathbb{R}) \rightarrow L^2(\mathbb{R})$:

$$g(t) = (Hf)(t) = \iint_{\mathbb{R} \times \mathbb{R}} \eta_H(\tau, \nu) \cdot f(t - \tau) e^{i2\pi\nu t} d\nu d\tau = \iint_{\mathbb{R} \times \mathbb{R}} \eta_H(\tau, \nu) (M_\nu T_\tau f)(t) d\nu d\tau$$

- ▷ Time discrete Time-Varying-Linear SISO Channel $H : L^2(\mathbb{R}) \rightarrow L^2(\mathbb{R})$:

$$g(t) = (Hf)(t) = \sum_{k=0}^{K-1} \sum_{\ell=0}^{M-1} \eta_H(k, \ell) \cdot f(t - k\Delta\tau) e^{i2\pi\ell\Delta\nu t} = \sum_{k=0}^{K-1} \sum_{\ell=0}^{M-1} \eta_H(k, \ell) \cdot f(t - kT) e^{i\frac{2\pi}{TL}\ell t}$$

- ▷ Rectification of the channel support region:
with $\Delta\tau = T$ and $\Delta\nu = \frac{1}{TL}$ for some prime $L \geq 5$.



Transmitted Signal

- ▷ **Transmitted signal:** Delta train followed by a guard interval

$$f(t) = \sum_{m=0}^{2L-1} x_m \delta(t - mT) \quad \text{with} \quad x_m = \begin{cases} \text{data symbol} & : 0 \leq m \leq L-1 \\ 0 & : L \leq m \leq 2L-1 \end{cases}$$

- ▷ **Received signal:**

1. Sampling at rate $1/T$: $g_n = (\mathbf{H}f)(nT), \quad n = 0, 1, \dots, 2L-1.$
2. Add two consecutive blocks of length L : $y_n = g_n + g_{n+L}, \quad n = 0, 1, \dots, L-1.$

- ▷ **Write in vector form:**

$$\mathbf{y} = \mathbf{H}\mathbf{x} = \sum_{\ell=0}^{L-1} \sum_{k=0}^{L-1} \eta(k, \ell) \mathbf{M}^{\ell} \mathbf{T}^k \mathbf{x} \quad (2)$$

with $\mathbf{x} = (x_0, \dots, x_{L-1})^T$ and $\mathbf{y} = (x_0, \dots, x_{L-1})^T$ and with $\mathbf{H} \in OPW(\Lambda).$

Remark: Using a periodic weighted delta train

$$f(t) = \sum_{p \in \mathbb{Z}} \sum_{m=0}^{2L-1} x_m \delta(t - mT - pLT)$$

results in a similar expression as in (2) but requires periodic data.

Transmission Scheme

- ▶ **Aim:** We want to **transmit a message** $\boldsymbol{\gamma} = \{\gamma_1, \dots, \gamma_Q\} \subset \mathbb{C}$ of size $Q \leq L$ over the channel $\mathbf{H} \in OPW(\Lambda)$ and recover $\boldsymbol{\gamma}$ at the receiver **without knowing \mathbf{H}** and **without knowing the channel support Λ** in advance.
- ▶ **Data symbols:** The data symbols a sum of a pilot and the actual message

$$\mathbf{x} = \mathbf{c} + \sum_{q=1}^Q \gamma_q \mathbf{e}_q$$

with the pilot signal \mathbf{c} which is chosen to be an *Alltop sequence* and \mathbf{e}_q are particular *chirp sequences*

$$\mathbf{c}(n) = \frac{1}{\sqrt{L}} \exp\left(i \frac{2\pi}{L} n^3\right) \quad \text{and} \quad \mathbf{e}_{mL+r}(n) = \frac{1}{\sqrt{L}} \exp\left(i \frac{2\pi}{L} [r + mn + rn^2]\right), \quad n \in \mathbb{Z}_L$$

- ▶ **Received signal:**

$$\mathbf{y} = \mathbf{H}\mathbf{x} = \sum_{\ell=0}^{L-1} \sum_{k=0}^{L-1} \eta(k, \ell) \mathbf{M}^{\ell} \mathbf{T}^k \mathbf{x} = \mathbf{G}(\mathbf{x})\boldsymbol{\eta} = \mathbf{G}(\mathbf{c})\boldsymbol{\eta} + \mathbf{U}\mathbf{s} = \boldsymbol{\Phi} \begin{pmatrix} \boldsymbol{\eta} \\ \mathbf{s} \end{pmatrix} \quad (3)$$

with a $L \times 2L^2$ measurement matrix $\boldsymbol{\Phi} = [\mathbf{G}(\mathbf{c}), \mathbf{U}]$ and a sparse vector $\mathbf{s} = f(\boldsymbol{\eta}, \boldsymbol{\gamma})$.

- ▶ **Recovery of the message:** Solve CS problem (3), then determine $\boldsymbol{\gamma} = g(\boldsymbol{\eta}, \mathbf{s})$.

Compressive Sampling Problem

▷ **Compressive Sampling Problem** of size $L \times 2L^2$

$$\mathbf{y} = \Phi \begin{pmatrix} \boldsymbol{\eta} \\ \mathbf{s} \end{pmatrix} \quad \text{with} \quad |\text{supp}(\boldsymbol{\eta}, \mathbf{s})^T| \leq (1 + Q) |\Lambda|$$

Lemma

The *coherence* of the measurement matrix $\Phi = [\mathbf{G}(\mathbf{c}), \mathbf{U}]$ is upper bounded by $\mu(\Phi) \leq \frac{2}{\sqrt{L}}$.

Remark: Welsh bound is $\frac{1}{\sqrt{L+1}}$.

Theorem

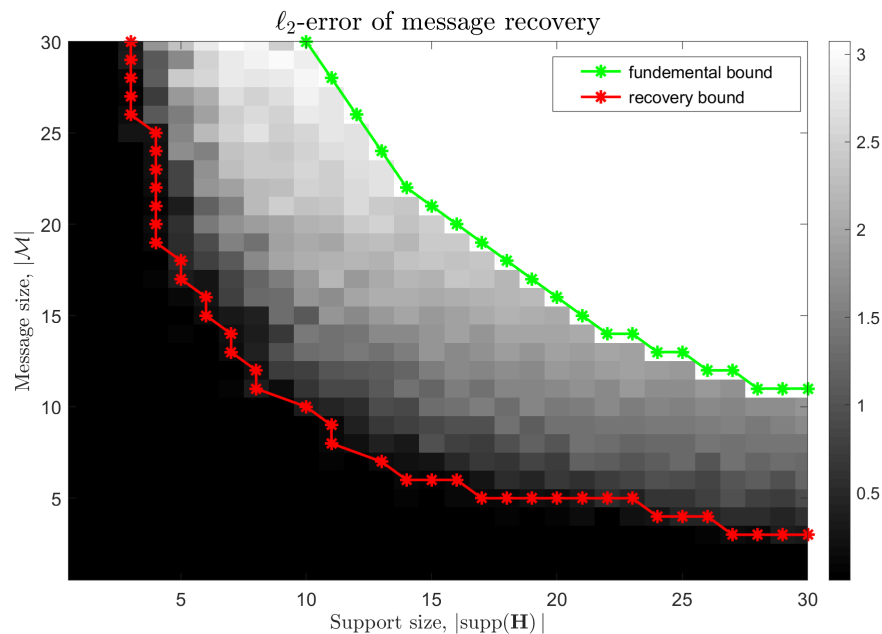
Let $\mathbf{H} \in \text{OPW}(\Lambda) \subset \mathcal{L}(\mathbb{C}^L)$ be an unknown channel with unknown support set Λ where $L \geq 5$ is a prime number. For

$$Q \leq \frac{\sqrt{L}}{4|\Lambda|} - 1$$

any message $\boldsymbol{\gamma} \in \mathbb{C}^Q$ can be transmitted over \mathbf{H} and recovered at the receiver.

Numerical Experiment

- $L = 307$
- OMP
- average over 100 channels (random support, Gaussian coefficients)
- red line: 1% rel. error
- simulation much better than bound



Summary

Current and Future Work






▷ Identification of SISO and MIMO TVL Channels:

- Identification of stochastic channels and stochastic sequences
- Conditions on support of the covariance of the spreading function η .
- Linear side constraints in terms of covariance.

▷ Transmission over Unidentified Channels:

- Stochastic encoding, RIP
- Scheme which maximum transmission rate
- Continuous time setting

Related Publications

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