

A Very Short Introduction to OTFS (Orthogonal Time Frequency Spacing)

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IERG5110 Course Presentation

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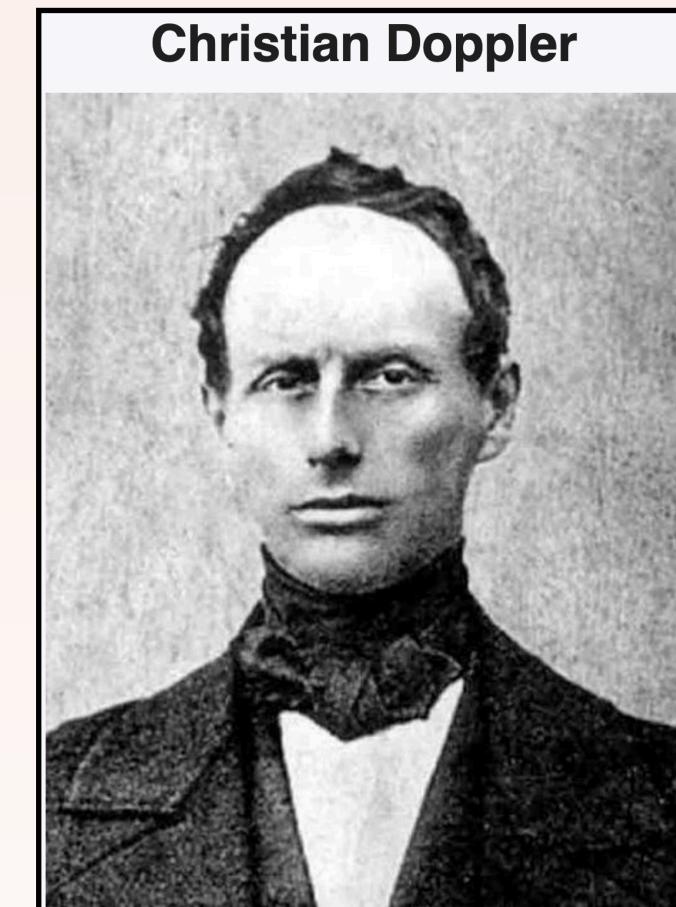
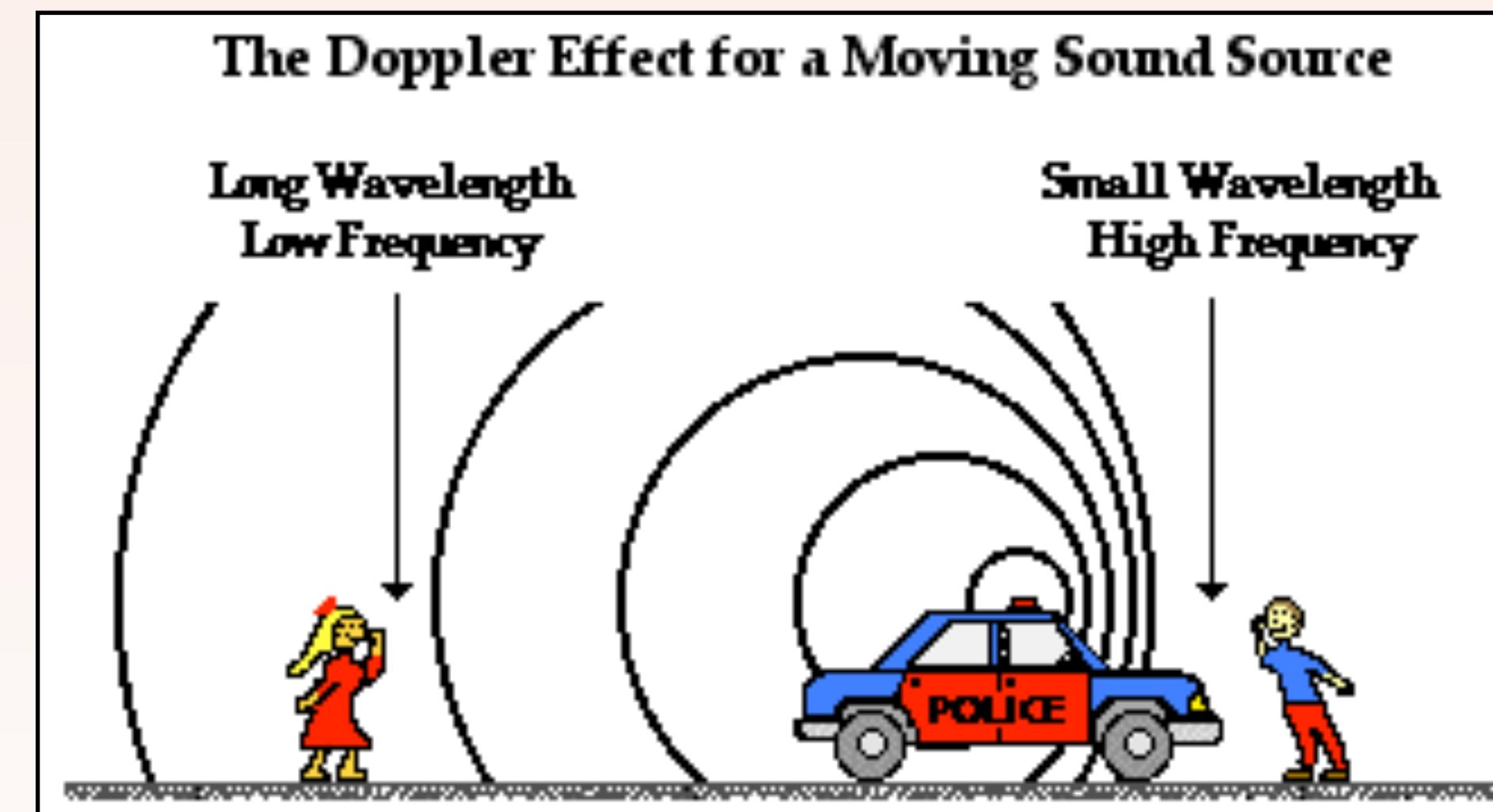


Content

- This < 1 hour talk will mainly focus on the basics of **OTFS (orthogonal time frequency spacing)**, and an inevitable wider concept: “**delay-Doppler communication/(de)modulation**”
 - Including: motivations, some derivations, the mathematical expressions, possibly with some examples.
- I tried to make this talk (and the slides as well) as simple & self-contained as they should be. So we can all learn something. Hopefully to develop some understanding.
- Also, as a humble learner on all the following topics, who knows little on this field. I have limited knowledge of the drawbacks and challenges associated with OTFS modulation.
- Specifically, we want to cover (not in the sequel):
 - **Motivation of OTFS (focus)**
 - **Must-know equations and expressions of OTFS (focus)**
 - **How to do channel estimation and message detection in DD domain (focus).**
 - Comparison between OTFS and OFDM
 - Some wonderful resources to learn OTFS
 - Code implementation for OTFS Tx-Rx and Mod/Demod (Toy example demo)

Delay and “Doppler” (OTFS modulation happens in the “delay-Doppler domain”/DD domain)

- What is delay? —A meeting is planned to start at **4pm**, but eventually started at **5pm**. This is called **delay**.
- In some sense, “delay” means **the amount of inconsistency / shift in the time domain**.

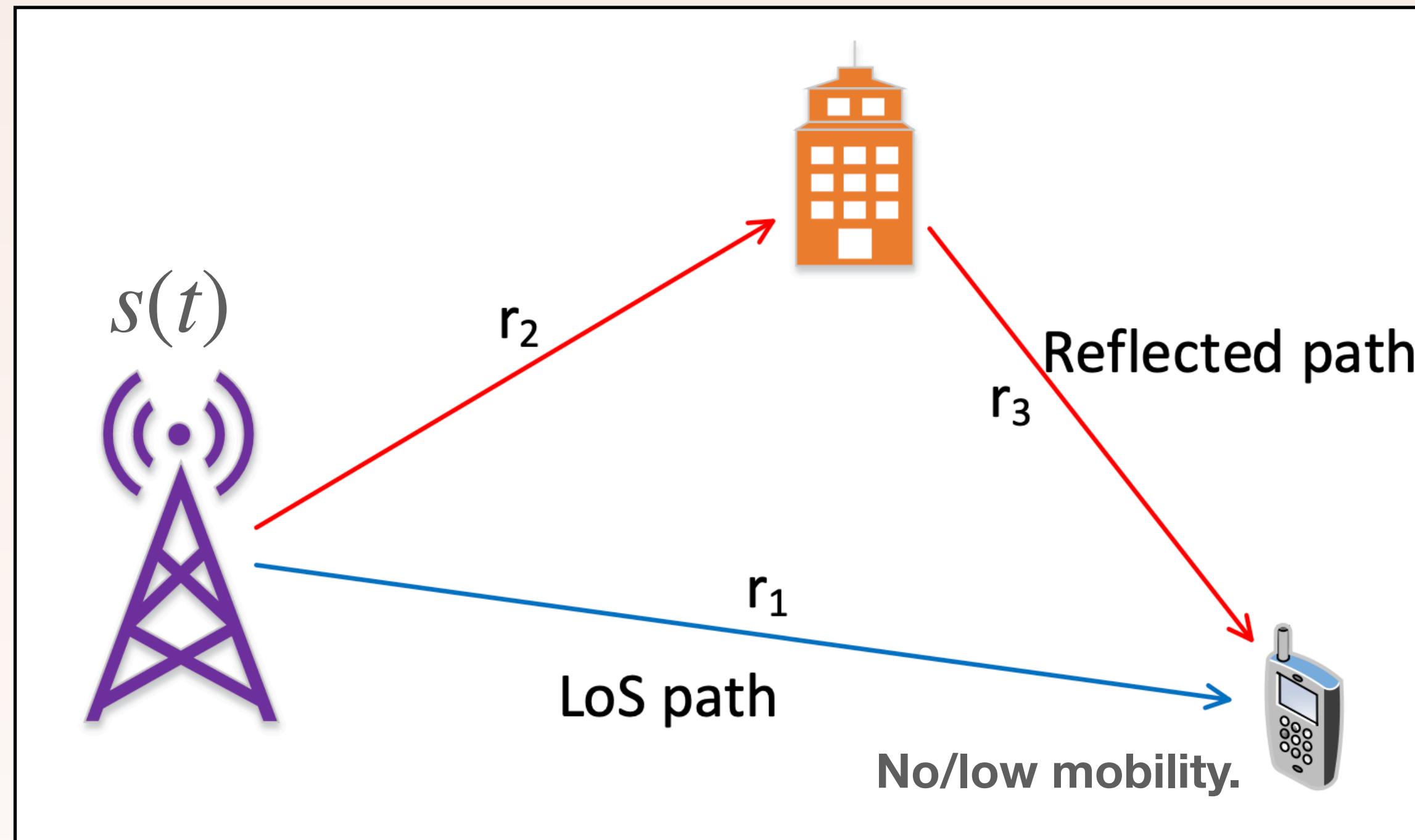


C. Doppler (1803 - 1853)

- What is “Doppler” (“Doppler shift”)?
 - — Doppler is a person. :)

- Something should be f_c Hz, but received as $f_c + f_d$ Hz. This additional amount f_d is termed as “**Doppler**”.
- “*A Doppler shift by f_d Hz is equivalent to modulating/upconverting the transmitted signal with $e^{j2\pi f_d t}$* ”.
- “Doppler” can be interpreted as: **the amount of inconsistency/ shift / difference in the frequency domain**.
- **What brings Doppler?** — **Mobility!**

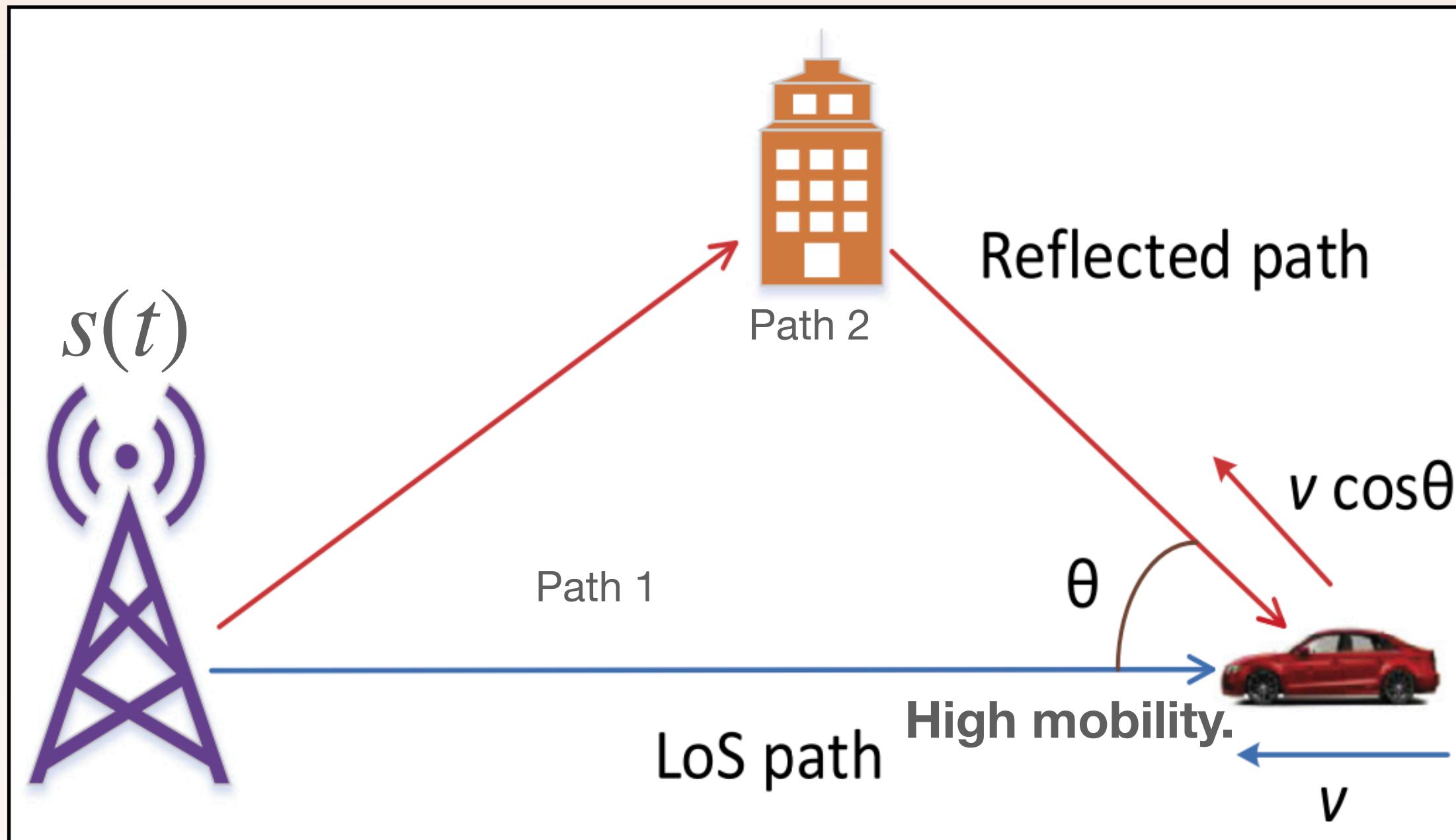
Low-mobility wireless channels



- Delay of LoS path: $\tau_1 = r_1/c$
- Delay of reflected path: $\tau_2 = (r_2 + r_3)/c$
- Delay spread: $\tau_2 - \tau_1$

- Suppose the signal sent by the BS is $s(t)$. Then what is received at the user (the phone) can be expressed as:
- $$r(t) = g_1 s(t - \tau_1) + g_2 s(t - \tau_2),$$
- where g_1, g_2 are (baseband equivalent) complex channel gains. And $c = 3 \cdot 10^8 m/s$ is the speed of light.

High-mobility wireless channels



- Hence, finally at the receiver, it receives (no noise yet)
- $r(t) = g_1 e^{j2\pi\nu_1(t-\tau_1)} s(t - \tau_1) + g_2 e^{j2\pi\nu_2(t-\tau_2)} s(t - \tau_2)$
- Note that, if there is no mobility \Rightarrow No Doppler $\Leftrightarrow \nu_i = 0 \Rightarrow$
- $g_1 s(t - \tau_1) + g_2 s(t - \tau_2)$ (What we wrote on the last slide)

- Original signal: $s(t)$
- On path i , $i = 1, 2$ considering the Doppler shift (an additional subcarrier): $s(t)e^{j2\pi\nu_i t}$
- In this example: $\nu_1 = \frac{v}{c} f_c$. And $\nu_2 = \frac{v \cos \theta}{c} f_c$.
- Consider delay: path i contribute $s(t - \tau_i)e^{j2\pi\nu_i(t-\tau_i)}$

	$f_c = 2 \text{ GHz}$	$f_c = 60 \text{ GHz}$
$v = 1.5 \text{ m/s} = 5.5 \text{ km/h}$	$\nu_{\max} = 10 \text{ Hz}$	$\nu_{\max} = 300 \text{ Hz}$
$v = 3 \text{ m/s} = 11 \text{ km/h}$	$\nu_{\max} = 20 \text{ Hz}$	$\nu_{\max} = 600 \text{ Hz}$
$v = 30 \text{ m/s} = 110 \text{ km/h}$	$\nu_{\max} = 200 \text{ Hz}$	$\nu_{\max} = 6 \text{ kHz}$
$v = 150 \text{ m/s} = 550 \text{ km/h}$	$\nu_{\max} = 1 \text{ kHz}$	$\nu_{\max} = 30 \text{ kHz}$

Doppler spread for some typical wireless channels.

Let take the last row as an example:

$$\nu_{\max} = \frac{v}{c} f_c = \frac{150}{3 \times 10^8} 2 \times 10^9 \text{ Hz} = 10^3 \text{ Hz} = 1 \text{ kHz}$$

Why OFDM is not perfect in high-mobility scenarios

- OFDM stacks its info in frequency domain: $x \in \mathbb{C}^{M \times 1}$. After a M -point FFT, it becomes the time domain OFDM (symbol) vector $\tilde{x} = \mathbf{F}_M^\dagger x \in \mathbb{C}^{M \times 1}$. (For rectangular pulse-shaping waveforms), we equip the CP part as:

- $s_{CP} = [\tilde{x}[M - l_{\max}], \dots, \tilde{x}[M - 1]]^T$, and obtain $s' = \underbrace{[s_{CP}^T, \tilde{x}^T]}_{\text{To transmit, length is } M + l_{\max}}$.

- Suppose the channel is **static**, ($\nu_i = 0$), then: $h(t) = \sum_{i=1}^P g_i \delta(t - \tau_i)$. Discrete-time equivalent channel vector:
 $h' \triangleq [h_0, \dots, h_{l_{\max}}]^T$. $h_l = g_i$ if $l = l_i, \forall i = 1, \dots, P$

- The equivalent baseband signal (including CP) is given by: (In time domain, convolution happens)

- $r' = \underbrace{h' * s'}_{\text{To receive}} + w' \in \mathbb{C}^{M+l_{\max}}$.

- Take away the CP, we have $r = [r'[l_{\max}], \dots, r'[M + l_{\max} - 1]]^T = h \circledast \tilde{x} + w$.
 $h = [h_0, \dots, h_{l_{\max}}, 0, \dots, 0]^T \in \overbrace{\mathbb{C}^M}^{\text{Removed CP, length is } M}$

- Equivalently, we can express $r = \mathbf{H} \tilde{x} + w$ for some appropriate matrix \mathbf{H} .

$\mathbf{H} =$

$$\begin{bmatrix} h_0 & 0 & \cdots & 0 & h_{l_{\max}} & h_{l_{\max}-1} & \cdots & h_1 \\ h_1 & h_0 & \cdots & 0 & 0 & h_{l_{\max}} & \cdots & h_2 \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ h_{l_{\max}-1} & \ddots & \ddots & \ddots & \ddots & \ddots & 0 & h_{l_{\max}} \\ h_{l_{\max}} & h_{l_{\max}-1} & \ddots & \ddots & \ddots & \ddots & \ddots & 0 \\ 0 & h_{l_{\max}} & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & h_0 & 0 \\ 0 & 0 & \cdots & h_{l_{\max}} & h_{l_{\max}-1} & \cdots & h_1 & h_0 \end{bmatrix},$$

$M \times M$ Circulant matrix

\mathbf{H} is a circulant matrix.

Why OFDM is not perfect in high-mobility scenarios

- Equivalently, we can express $r = \mathbf{H}\tilde{x} + w$

$$\mathbf{H} = \sum_{i=1}^P g_i \Pi^{l_i}$$

$$\mathbf{H} = \begin{bmatrix} h_0 & 0 & \cdots & 0 & h_{l_{\max}} & h_{l_{\max}-1} & \cdots & h_1 \\ h_1 & h_0 & \cdots & 0 & 0 & h_{l_{\max}} & \cdots & h_2 \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ h_{l_{\max}-1} & \ddots & \ddots & \ddots & \ddots & 0 & h_{l_{\max}} & \\ h_{l_{\max}} & h_{l_{\max}-1} & \ddots & \ddots & \ddots & \ddots & 0 & \\ 0 & h_{l_{\max}} & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \\ 0 & 0 & \cdots & h_{l_{\max}} & h_{l_{\max}-1} & \cdots & h_1 & h_0 \end{bmatrix},$$

M × M Circulant matrix

Static channel (we saw on the last page)

- Note that on the LHS, \mathbf{H} is a circulant matrix: which can be diagonalized via Fourier transform matrices pairs (DFT matrices form a orthonormal basis of circulant matrices)

- $\mathbf{H} = \mathbf{F}_M^\dagger \text{diag}(\mathbf{F}_M h) \mathbf{F}_M$.

- $y = \mathbf{F}_M r = \hat{h} \odot \tilde{x} + \hat{w}$

- Where \odot denotes element-wise product. Such orthogonality enables us to use **single-tap equalizer** to estimate \tilde{x} entry-by-entry.

$$\boldsymbol{\Pi} = \begin{bmatrix} 0 & \cdots & 0 & 1 \\ 1 & \ddots & 0 & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & \cdots & 1 & 0 \end{bmatrix}$$

$$\mathbf{H} = \sum_{i=1}^P g_i \Pi^{l_i} \mathbf{D}^{k_i}, \quad \mathbf{D} \triangleq \text{diag}(1, \omega, \dots, \omega^{M-1})$$

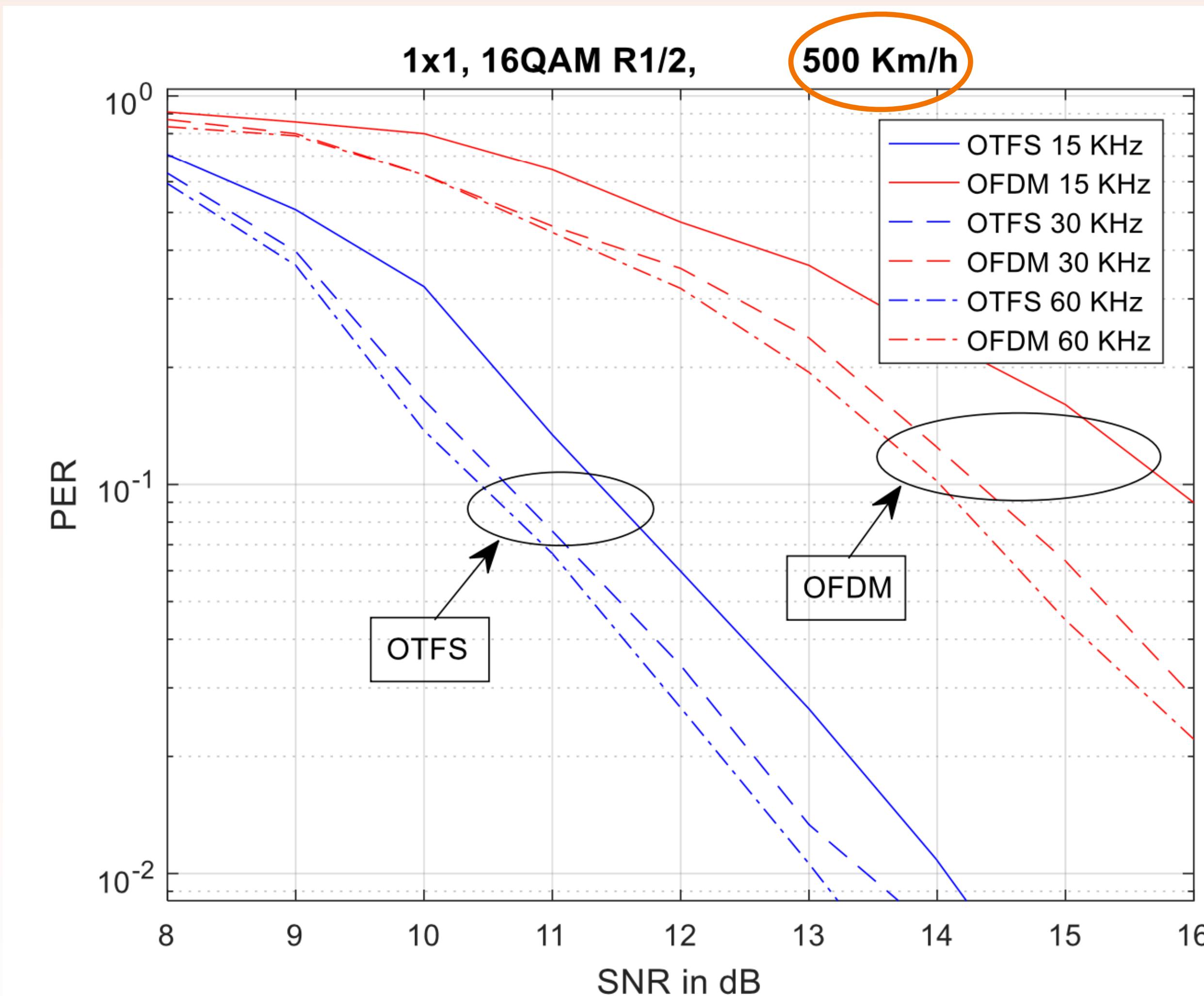
- How to understand this? —— Each path $i \in [P]$. The signal is first “modulated” by a carrier k_i . Then shifted by Π^{l_i} . Then they composite together to make the received r , with noise.

$$= \begin{bmatrix} h_0 & 0 & \cdots & h_{l_{\max}} \omega^{k_P(M-l_{\max})} & \cdots & h_1 \omega^{k_1(M-1)} \\ \vdots & h_0 \omega^{k_0} & \cdots & \ddots & \cdots & \vdots \\ \vdots & \ddots & \ddots & \ddots & 0 & \ddots \\ h_{l_{\max}} & \ddots & \ddots & \ddots & \ddots & \ddots \\ 0 & h_{l_{\max}} \omega^{k_P} & \ddots & \ddots & \ddots & \ddots \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots \\ 0 & \cdots & h_{l_{\max}} \omega^{k_P(M-l_{\max}-1)} & \cdots & \cdots & h_0 \omega^{k_0(M-1)} \end{bmatrix}$$

High-mobility channel

- However, on the RHS, there are some additional ω^{\cdots} factors, caused by Doppler shifts, makes this matrix no longer circulant. \Rightarrow **cannot be eigen-decomposed via FFT matrices!!!** (Where $\omega = e^{\frac{j2\pi}{M}}$)
- Continuous case: $s(t) \rightarrow s(t)e^{j2\pi f_m t}$. Discrete case: $s[m] \rightarrow s[m]e^{j2\pi k_m}$
- \Rightarrow Gives us performance degradation. If we do LHS decoding anyway \Rightarrow need to endure some ICI.
If those $k_m = 0$, then RHS \rightarrow LHS.

With **Dopper**, OTFS outperforms OFDM



This plot should be self-explanatory.

High-mobility wireless channels

- The 2-path result can be generalized for P paths as the following:

$$r(t) = \sum_{i=1}^P g_i e^{j2\pi\nu_i(t-\tau_i)} s(t - \tau_i) = \sum_{i=1}^P g_i e^{j2\pi\nu_i t} s(t - \tau_i) e^{-j2\pi\nu_i \tau_i}$$

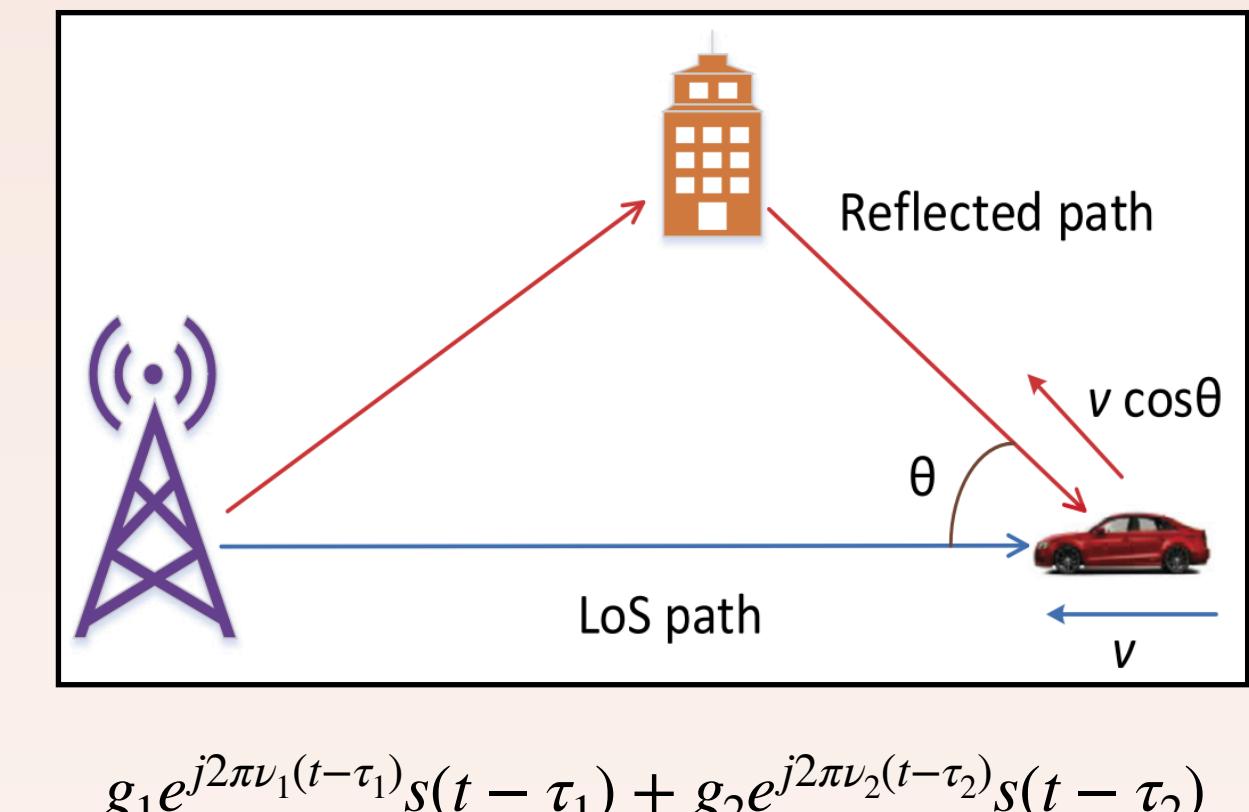
- We define the delay-Doppler channel representation as:

$$h(\tau, \nu) \triangleq \sum_{i=1}^P g_i e^{-j2\pi\nu_i \tau_i} \delta(\tau - \tau_i) \delta(\nu - \nu_i) \quad [\text{this function only has value on } (\tau_i, \nu_i)]$$

- One can see the received signal $r(t)$ as:

$$r(t) = \int \int h(\tau, \nu) e^{j2\pi\nu t} s(t - \tau) d\nu d\tau$$

- The cool thing here is:** the channel $h(\tau, \nu)$ (in DD domain) is completely represented by these quantities: $(g_i, \tau_i, \nu_i), i = 1, 2, \dots, P$. The # of quantities you need is small \implies sparsity.
- We will relate the above new channel representation with something we knew.



Different (Linear time variant) channel representations

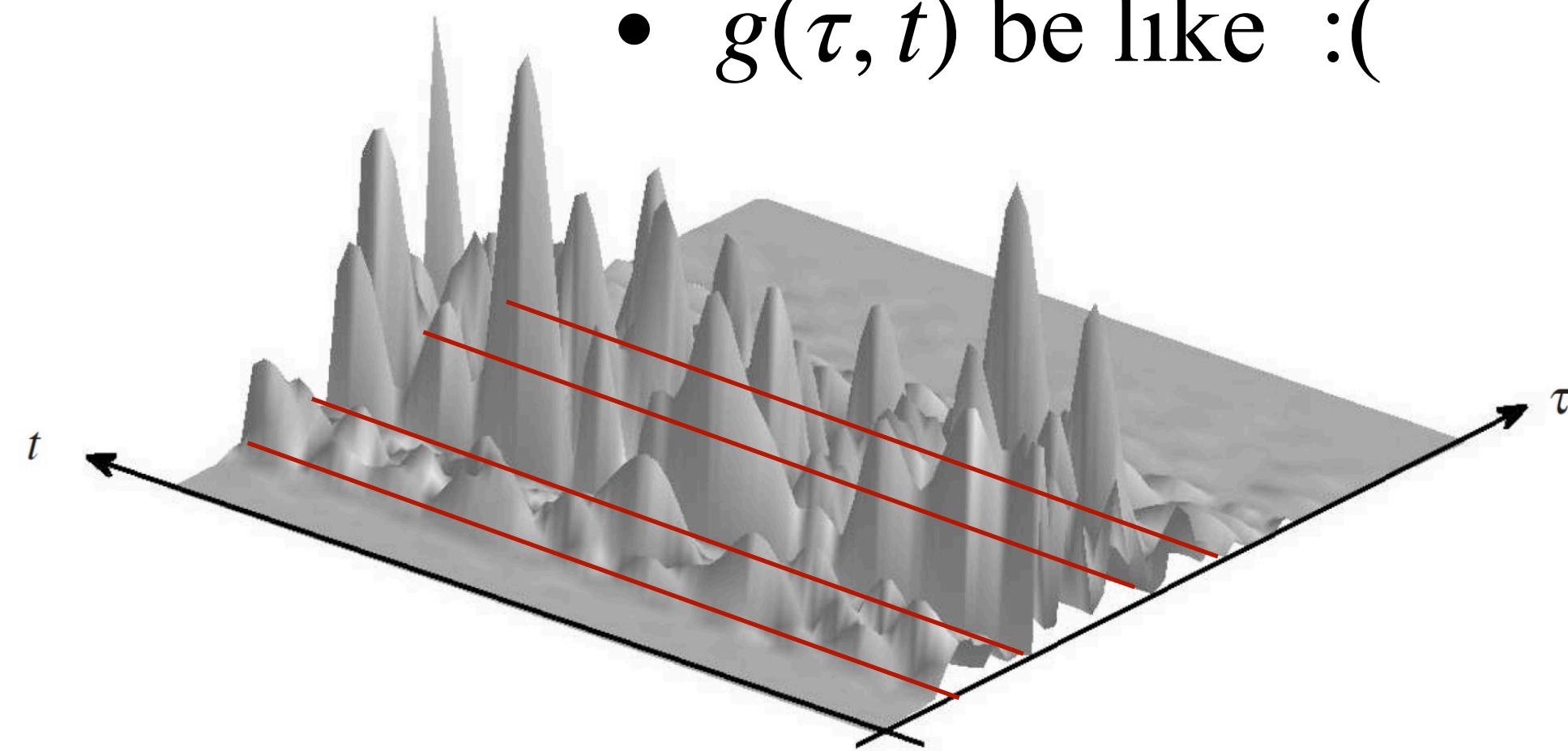
- $$h(\tau, \nu) \triangleq \sum_{i=1}^P g_i e^{-2\pi\nu_i \tau_i} \delta(\tau - \tau_i) \delta(\nu - \nu_i) \quad (\text{delay-Doppler})$$

This is the delay-time domain representation of the channel.

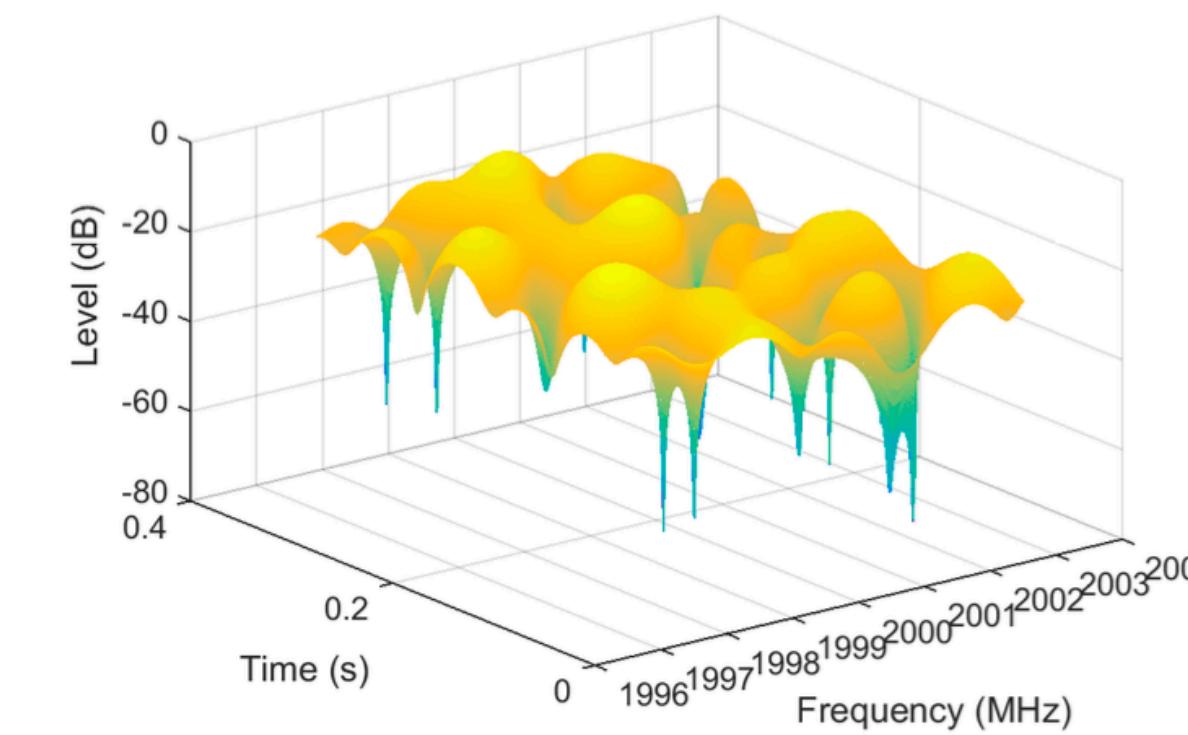
- If we go back one step: $r(t) = \sum_{i=1}^P g_i e^{j2\pi\nu_i(t-\tau_i)} s(t - \tau_i)$
- $r(t) = \sum_{i=1}^P \underbrace{g_i e^{j2\pi\nu_i(t-\tau_i)}}_{g(\tau_i, t)} s(t - \tau_i) \implies \text{generalize } \int_0^\infty \underbrace{g(\tau, t)}_{\text{Delay-time domain}} s(t - \tau) d\tau.$
- Another immediate equivalent representation is the time-frequency representation:
- $r(t) = \sum_{i=1}^P \underbrace{g_i e^{j2\pi\nu_i(t-\tau_i)}}_{g(\tau_i, t)} s(t - \tau_i) \implies \int_0^\infty g(\tau, t) s(t - \tau) d\tau = \int \underbrace{H(f, t) S(f)}_{\text{Time-frequency domain}} e^{j2\pi f t} df$
- The last step can be understood as consider product in frequency domain, then use IFT back to time domain.

Time-variant impulse response $g(t, \tau)$

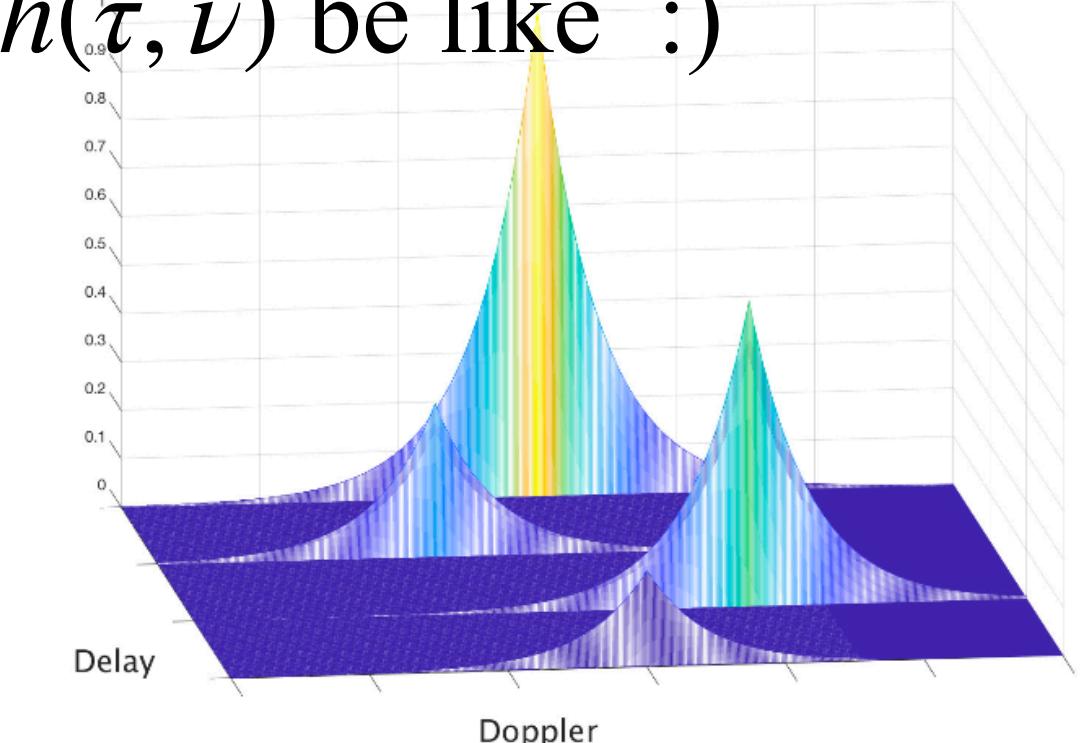
- $g(\tau, t)$ be like :(



- $h(\tau, \nu)$ be like :(

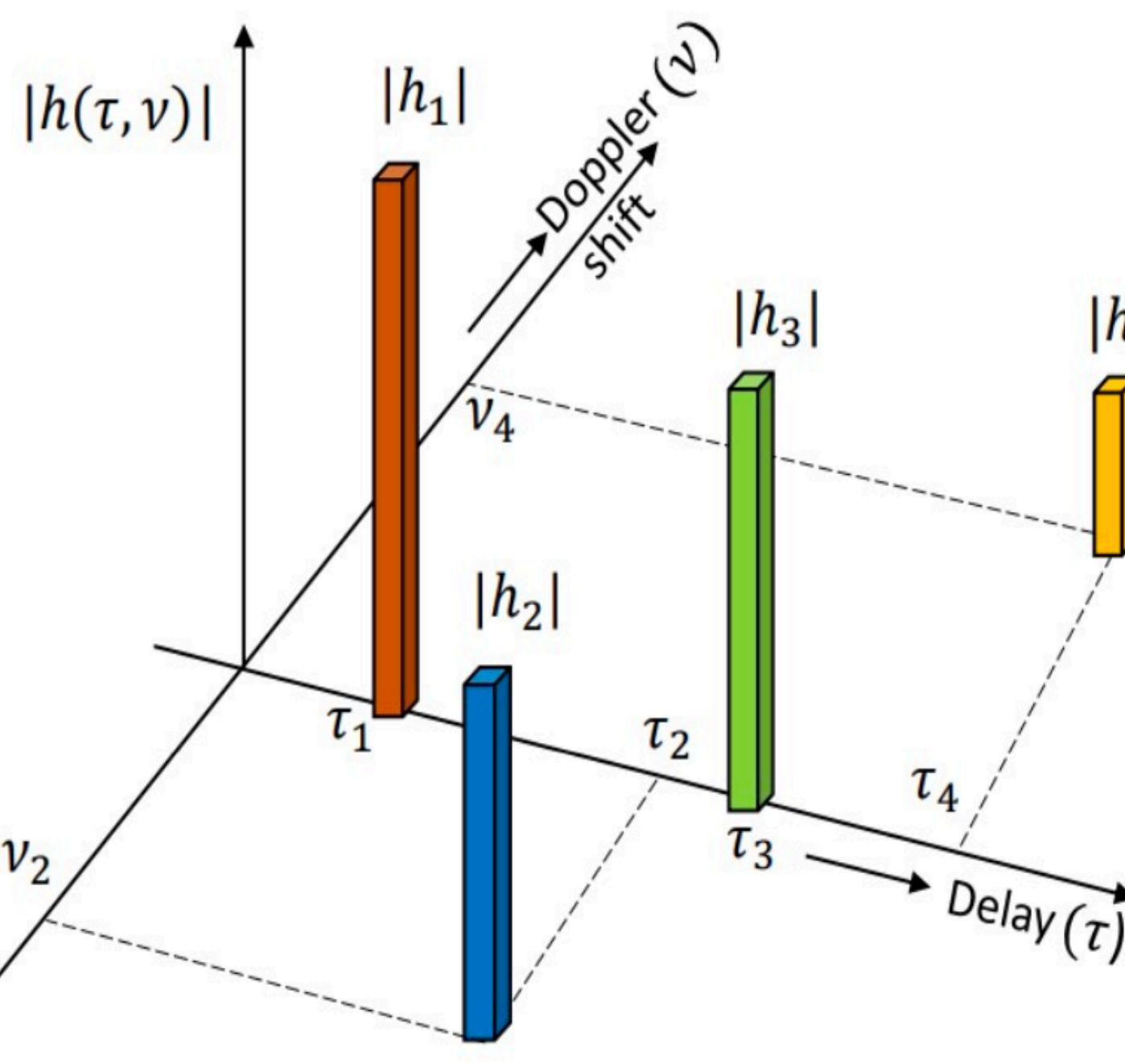
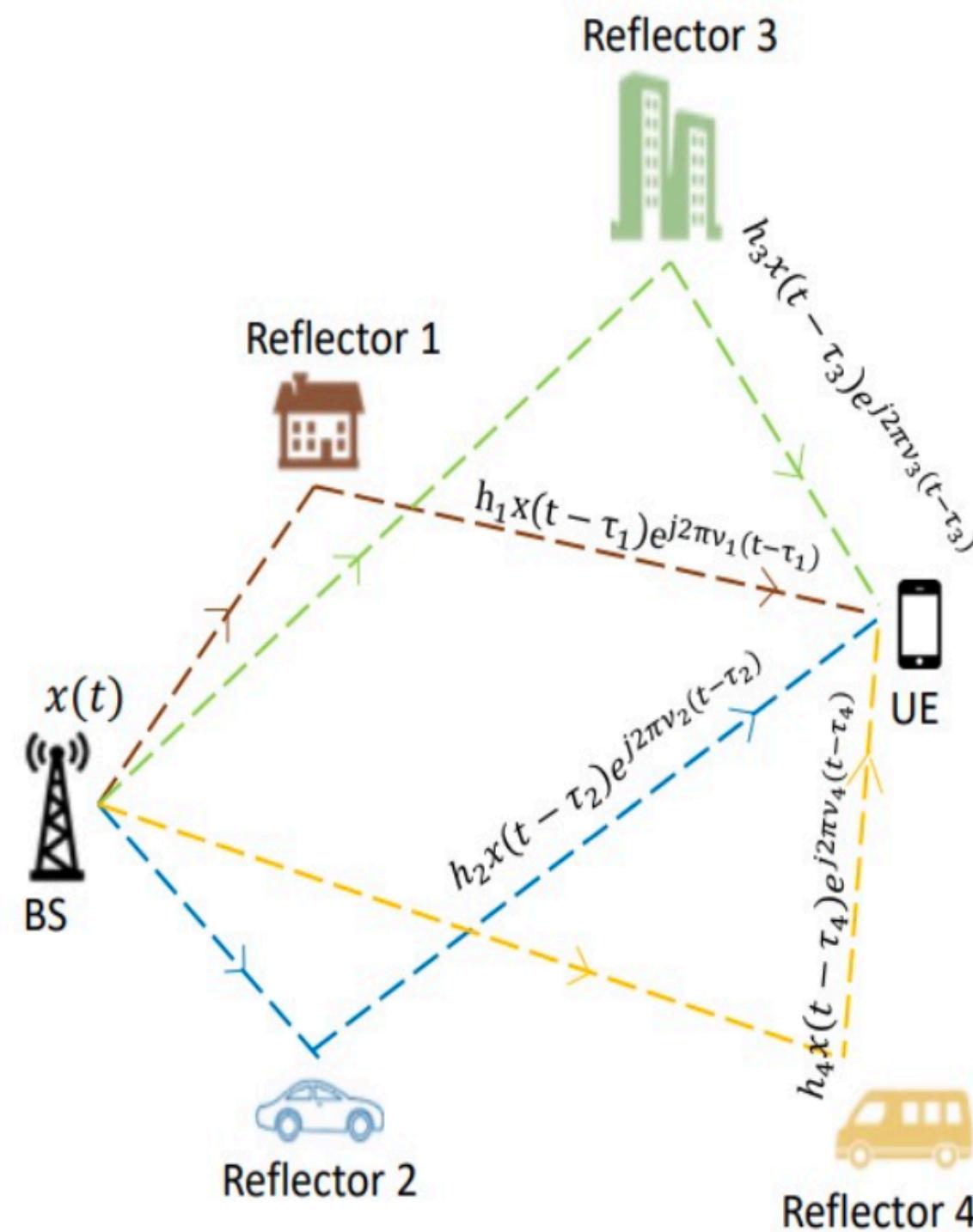


SFFT
ISFFT



- $H(t, f)$ be like :(

Channel in Time–frequency $H(t, f)$ and delay–Doppler $h(\tau, \nu)$

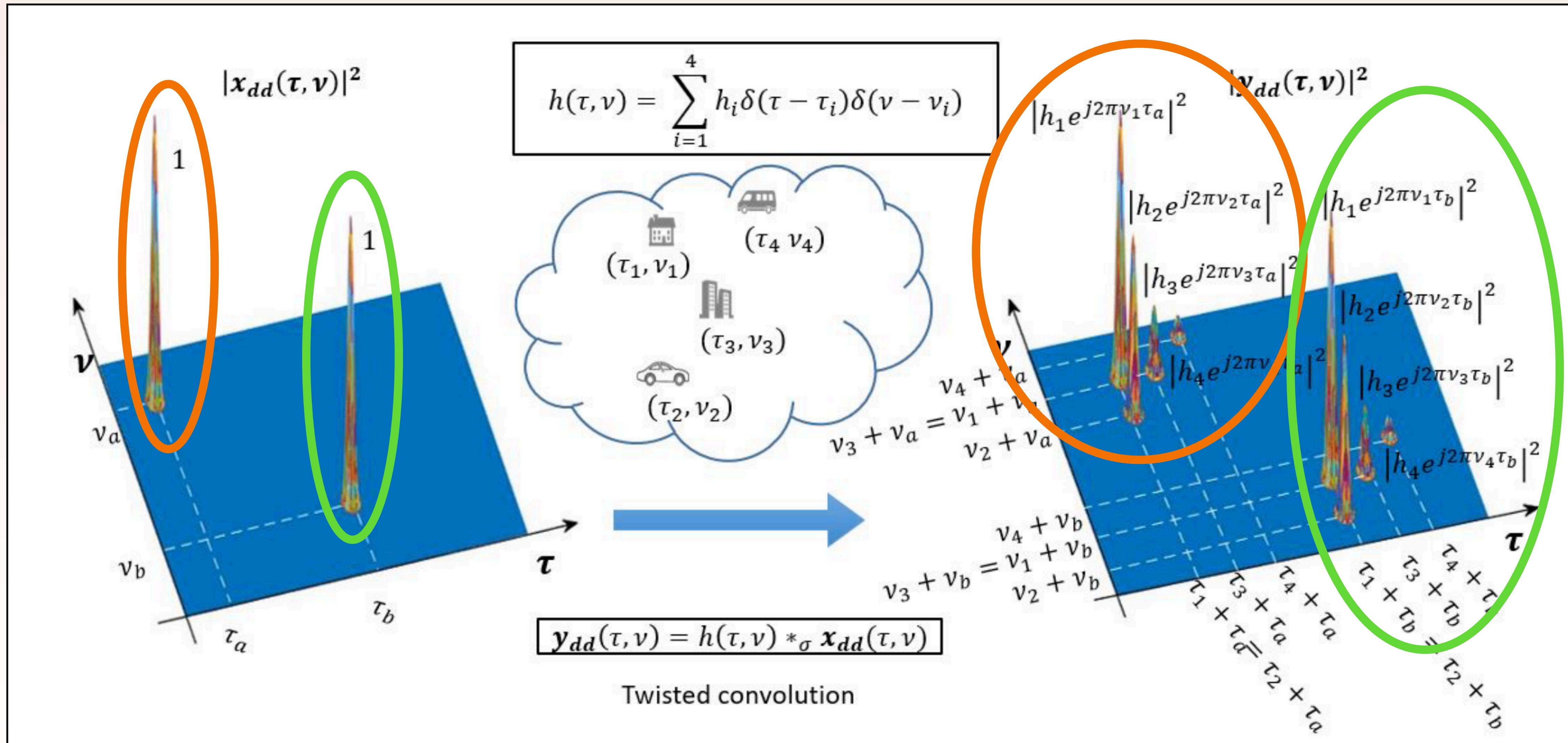


- DD domain notation sparsify the channel representation!!

- The channel representation should be simple as this!!!

TX and RX signal in *delay-Dopper domain*

- OK. But does the DD domain stuff ever make our lives easier?? —— Probably, yes.



- We will come back to this later. Now just get some intuitive feeling of how signals will interact with the channel in DD domain.

Parameters and their choice in OTFS

- In an OTFS system, we need to specify a bunch of parameters.
- We assume the OTFS system operates on a P -path high mobility channel, with a bandwidth B .
- Maximum delay spread τ_{\max} , maximum Doppler shift ν_{\max} .
- Sampling frequency $f_s = B = \frac{1}{T_s}$, where T_s is referred as “sampling interval”.
- Each OTFS frame contains NM samples divided into N time slots. M samples per block.
- OTFS frame duration $T_f = NMT_s = NT$. $T = MT_S$ denotes the duration of each block.
- $\Delta f = \frac{1}{T}$. This is the “space of spectrum samples”.

OTFS modulation

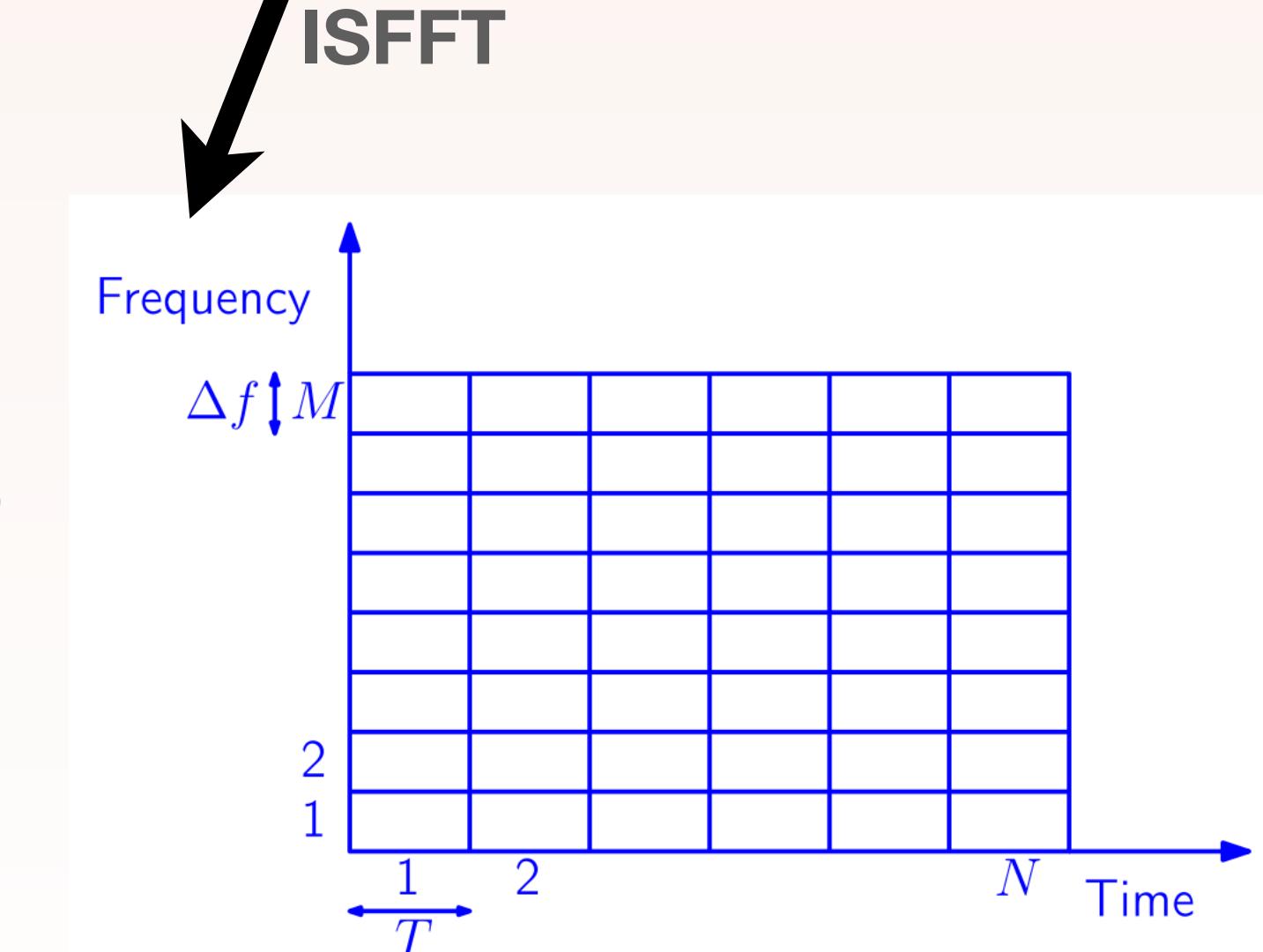
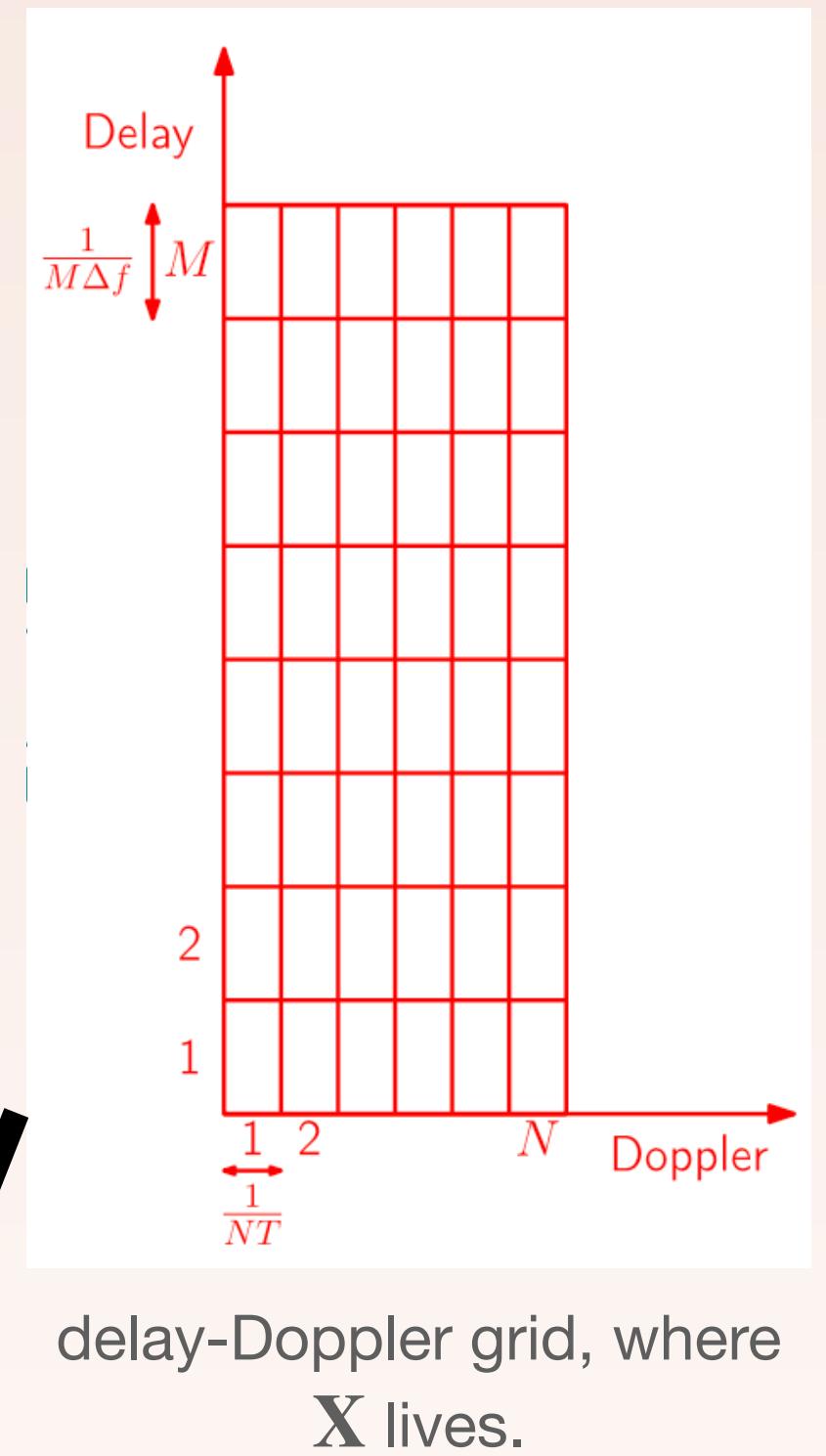
- In OFDM, we store our information in frequency domain vectors. In OTFS, we place them in “**delay-Doppler**” domain matrix $\mathbf{X} \in \mathbb{C}^{M \times N}$, with entries $\mathbf{X}[m, n], m \in [M], n \in [N]$.
- We first convert \mathbf{X} into frequency-time domain $\mathbf{X}_{ft} \in \mathbb{C}^{M \times N}$ by applying (N-pt) IFFT along the rows of \mathbf{X} , then apply (M-pt) FFT on columns of it. (Can we swap the order? —— yes) Put it in math, we have:

$$\mathbf{X}_{ft} = \frac{1}{\sqrt{MN}} \sum_{n=0}^{N-1} \sum_{m=0}^{M-1} \mathbf{X}[m, n] e^{j2\pi(\frac{nk}{N} - \frac{ml}{M})} = \mathbf{F}_M \underbrace{\mathbf{X}}_{M \times N} \mathbf{F}_N^H \quad (\text{this is called “ISFFT” inverse symplectic FFT})$$

- Then to convert the 2D \mathbf{X}_{ft} to a time domain signal, we do the following:

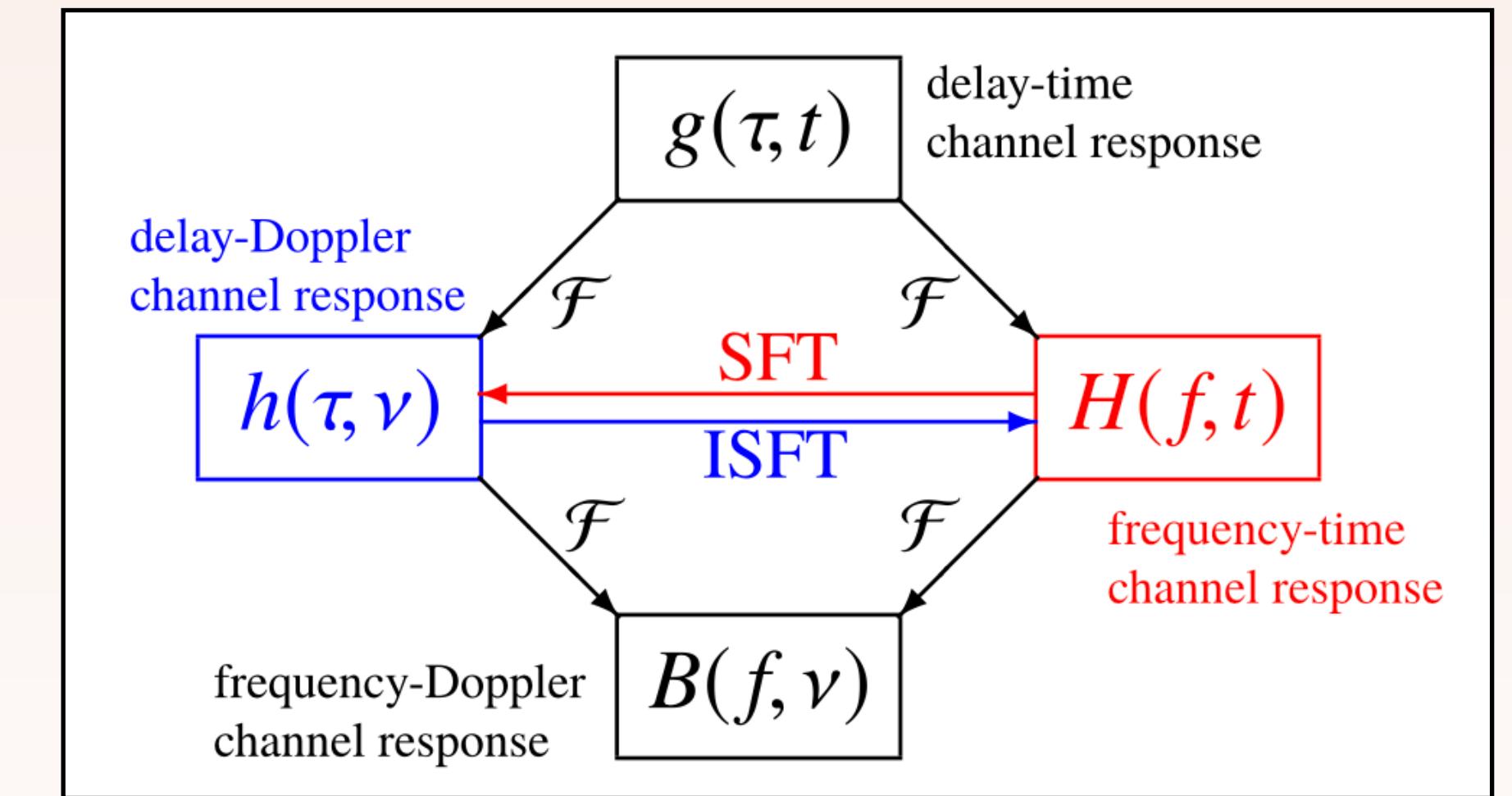
$$s(t) = \sum_{k=0}^{N-1} \sum_{l=0}^{M-1} \mathbf{X}_{ft}[l, k] \underbrace{g_{tx}(t - kT)}_{\text{Some transmit waveform}} e^{j2\pi l \Delta f (t - kT)} \quad (\text{this is called Heisenberg transform})$$

- The above $s(t)$ is ready to transmit.



OTFS: channel (revisited)

- $$s(t) = \sum_{k=0}^{N-1} \sum_{l=0}^{M-1} \mathbf{X}_{ft}[l, k] g_{tx}(t - kT) e^{j2\pi l \Delta f (t - kT)} \quad (\text{what to transmit})$$
- Impaired by the channel, we can express the received signal as the following:
- $$r(t) = \iint h(\tau, \nu) e^{j2\pi\nu t} s(t - \tau) d\nu d\tau \quad (\text{delay-Doppler channel}).$$
- $$= \int \underbrace{H(t, f)}_{\text{time-frequency response}} S(f) e^{j2\pi ft} df \quad (\text{Time-frequency channel}).$$
- $$= \int g(t, \tau) s(t - \tau) d\tau \quad [\text{delay - time expression, should be the most intuitive}]$$



Relations between different representations of LTV channels

OTFS: demodulation

- $s(t) = \sum_{k=0}^{N-1} \sum_{l=0}^{M-1} \mathbf{X}_{ft}[l, k] g_{tx}(t - kT) e^{j2\pi l \Delta f(t - kT)}, \quad r(t) = \int \int h(\tau, \nu) e^{j2\pi \nu t} s(t - \tau) d\nu d\tau$ (other expressions available)
- With a matched filter, we process like the following: Explanation needed
- $Y(f, t) = A_{g_{rx}, r}(f, t) \triangleq \int g_{rx}^*(t' - t) r(t') e^{-j2\pi f(t' - t)} dt'$ (This is called “Wigner transform”)
- “Cross-ambiguity function” between g_1, g_2 is defined as $A_{g_1, g_2}(f, t) \triangleq \int g_1(t') g_2^*(t' - t) e^{-j2\pi f(t' - t)} dt'$.
- Intuition? $A_{g_1, g_2}(f, t)$ defines the correlation between $g_1(t)$ and a delayed-by- t & frequency shifted by f version of $g_2(t)$.
- $\mathbf{Y}_{ft}[l, k] = Y(f, t) |_{f=l\Delta f, t=kT} \in \mathbb{C}^{M \times N}$ This lives in frequency-time domain
- Finally, we use the SFFT to process $\mathbf{Y}_{ft}[l, k]$, leading to a matrix \mathbf{Y} :
- $\mathbf{Y}[m, n] = \frac{1}{\sqrt{NM}} \sum_{k=0}^{N-1} \sum_{l=0}^{M-1} \mathbf{Y}_{ft}[l, k] e^{-j2\pi(\frac{nk}{N} - \frac{ml}{M})}$ This lives in DD domain
- *This finishes the “standard procedure” of OTFS modulation & demodulation. (But understand little 😞)*

Delay-Doppler domain analysis

- The delay-Doppler transmitted and received samples matrices are related to those in the frequency-time domain by SFFT:

$$\mathbf{Y} = \text{SFFT}(\mathbf{Y}_{ft}), \quad \mathbf{X} = \text{SFFT}(\mathbf{X}_{ft}).$$

- Also, the delay-Doppler domain channel matrix (of size $M \times N$) is related to the frequency-time channel matrix as:

$$\mathbf{H}_{dd} = \text{SFFT}(\mathbf{H}_{ft}).$$

- (In frequency-time domain, ideally) $\mathbf{Y}_{ft}[l, k] = \mathbf{H}_{ft}[l, k]\mathbf{X}_{ft}[l, k].$ ➡ Omit/fix the k , nothing but $Y(f) = H(f)S(f)$

- Then we can actually derive:

$$\mathbf{Y}[m, n] = \mathbf{H}_{dd}[m, n] \circledast \mathbf{X}[m, n]$$

$$= \sum_{m'} \sum_{n'} \mathbf{H}_{dd}[m', n'] \mathbf{X}[[m - m']_M, [n - n']_N]$$

$$= \sum_{i=1}^P g_i \mathbf{X}[[m - l_i]_M, [n - k_i]_N]$$

- Lemma: $\mathbf{Z} = \mathbf{U} \circledast \mathbf{V}$ can be equivalently written as $\text{vec}(\mathbf{Z}) = \mathbf{B}_U \cdot \text{vec}(\mathbf{V})$, where $\mathbf{B}_U = \text{circ}[\text{circ}(U_1), \dots, \text{circ}(U_N)].$

$$\text{circ}(a_0, \dots, a_{N-1}) = \begin{pmatrix} a_0 & a_{N-1} & \cdots & & a_1 \\ a_1 & a_0 & & & a_2 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ a_{N-2} & & & a_0 & a_{N-1} \\ a_{N-1} & a_{N-2} & \cdots & a_1 & a_0 \end{pmatrix}$$

Delay-Doppler domain analysis

- (In frequency-time domain, ideally) $\mathbf{Y}_{ft}[l, k] = \mathbf{H}_{ft}[l, k]\mathbf{X}_{ft}[l, k]$.

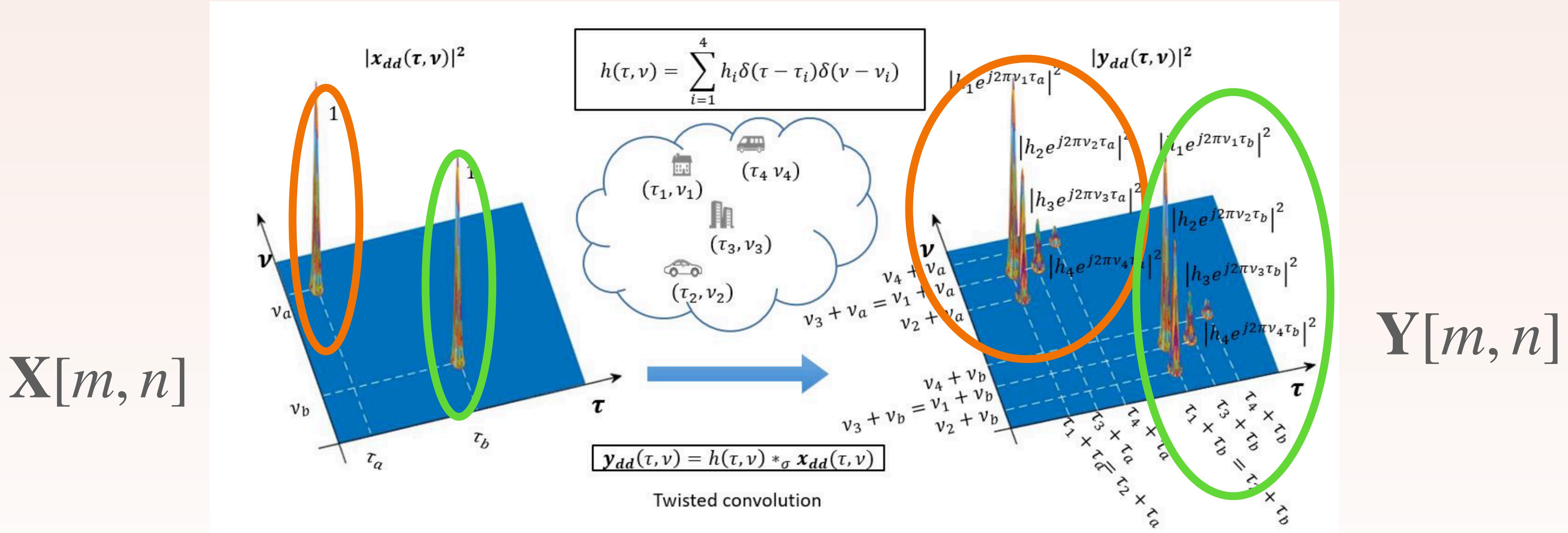
⬇️ want to derive the bottom

- $\mathbf{Y}[m, n] = \mathbf{H}_{dd}[m, n] \circledast \mathbf{X}[m, n] \triangleq \sum_{m'} \sum_{n'} \mathbf{H}_{dd}[m', n'] \mathbf{X}[[m - m']_M, [n - n']_N]$.

- Proof sketch (details is quite tedious, if we do not use higher level theorems)

- Suffice to prove $\text{vec}(\mathbf{Y}) = \mathbf{B}_{\mathbf{H}_{dd}} \cdot \text{vec}(\mathbf{X})$
- $\mathbf{Y}_{ft}[l, k] = \mathbf{H}_{ft}[l, k]\mathbf{X}_{ft}[l, k] \iff \text{vec}(\mathbf{Y}_{ft}) = \text{vec}(\mathbf{H}_{ft}) \odot \text{vec}(\mathbf{X}_{ft}) \quad \odot \text{ element-wise / Hadarmard product}$
- $\iff (\mathbf{F}_N \otimes \mathbf{F}_M^\dagger)^{-1} (\mathbf{F}_N \otimes \mathbf{F}_M^\dagger) \text{vec}(\mathbf{Y}_{ft}) = \text{diag}(\text{vec}(\mathbf{H}_{ft})) (\mathbf{F}_N \otimes \mathbf{F}_M^\dagger)^{-1} (\mathbf{F}_N \otimes \mathbf{F}_M^\dagger) \text{vec}(\mathbf{X}_{ft})$
 - ⊗ Kronecker product $\underbrace{\mathbf{F}_N \otimes \mathbf{F}_M^\dagger}_{=\text{vec}(\mathbf{Y})}$ $\underbrace{\mathbf{F}_N \otimes \mathbf{F}_M^\dagger}_{=\text{vec}(\mathbf{X})}$
- $\iff \text{vec}(\mathbf{Y}) = (\mathbf{F}_N \otimes \mathbf{F}_M^\dagger) \underbrace{\text{diag}(\text{vec}(\mathbf{H}_{ft})) (\mathbf{F}_N \otimes \mathbf{F}_M^\dagger)^{-1}}_{\text{This, surprisingly, equals } \mathbf{B}_{\mathbf{H}_{dd}}} \text{vec}(\mathbf{X})$

TX and RX signal in DD domain (revisited)



$$Y[m, n] = H_{dd}[m, n] \circledast X[m, n]$$

OTFS: modulation & demodulation summary

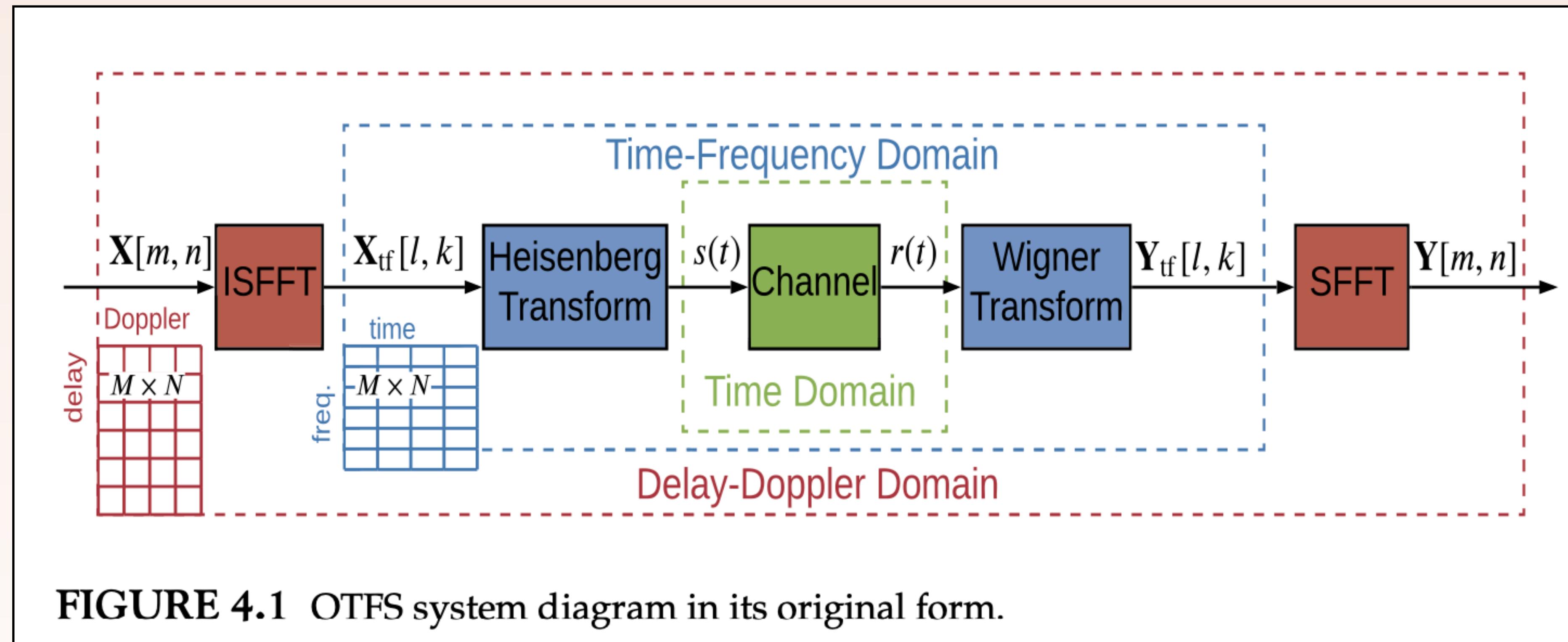


FIGURE 4.1 OTFS system diagram in its original form.

- As shown above, the OTFS modulator maps $\mathbf{X}[m, n]$ in the delay-Doppler domain to $\mathbf{X}_{tf}[l, k]$ in the time-frequency domain (using ISFFT). Then the Heisenberg transform is applied to $\mathbf{X}_{tf}[l, k]$ to generate time domain signal $s(t)$.
- At the receiver, $r(t)$ is transformed to the time-frequency domain by Wigner transform. Then to DD domain via SFFT.

Channel estimation (in delay-Doppler way)

- Estimation in DD domain. (There are other domain-based channel estimation methods.)

$N - 1$	x	x	x	x	x	x	x	x	x	x	x
	x	x	x	x	x	x	x	x	x	x	x
$k_p + 2k_\nu$	x	x	x	o	o	o	o	x	x	x	x
	x	x	x	o	o	o	o	x	x	x	x
k_p	x	x	x	o	o	o	o	x	x	x	x
	x	x	x	o	o	o	o	x	x	x	x
$k_p - 2k_\nu$	x	x	x	o	o	o	o	x	x	x	x
1	x	x	x	x	x	x	x	x	x	x	x
0	x	x	x	x	x	x	x	x	x	x	x
	0	1	$l_p - l_\tau$	l_p	$l_p + l_\tau$	$M - 1$					

(a) Tx symbol arrangement (\square : pilot; \circ : guard symbols;
 \times : data symbols)

Zero symbols

$N - 1$	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇
	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇
$k_p + k_\nu$	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇
k_p	∇	∇	∇	∇	∇	\blacksquare	\blacksquare	\blacksquare	\blacksquare	∇	∇
$k_p - k_\nu$	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇
1	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇
0	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇	∇
	0	1	$l_p - l_\tau$	l_p	$l_p + l_\tau$	$M - 1$					

(b) Rx symbol pattern (∇ : data detection, \blacksquare : channel estimation)

With channel estimation, how to decode?

- Suppose one have estimated the (delay-Doppler domain) channel matrix $\mathbf{H} \in \mathbb{C}^{NM \times NM}$, then we are facing (equivalently) a linear system with NM equations:
- $\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{w}$, (Comes from $\mathbf{Y} = \mathbf{H}_{dd} \circledast \mathbf{X}$, all $M \times N$ matices, using the previous lemma)
- Where where $\mathbf{y}, \mathbf{x}, \mathbf{w} \in \mathbb{C}^{NM \times 1}$ are $\mathbf{x}[k, l], \mathbf{y}[k, l], \mathbf{w}[k, l]$ in vectorized form, $k \in [N], l \in [M]$. \mathbf{H} is treated as given.
- Given the sparse nature of \mathbf{H} (each row sum=each column sum = P) we can solve it by message passing algorithm.

$$\widehat{\mathbf{x}} = \arg \max_{\mathbf{x} \in \mathbb{A}^{NM \times 1}} \Pr(\mathbf{x} \mid \mathbf{y}, \mathbf{H}),$$

which has a complexity exponential in NM . Since the joint MAP detection can be intractable for practical values of N and M , we consider the symbol-by-symbol MAP detection rule for $c = 1, \dots, NM$

$$\widehat{x}[c] = \arg \max_{a_j \in \mathbb{A}} \Pr(x[c] = a_j \mid \mathbf{y}, \mathbf{H}) \quad (31)$$

$$= \arg \max_{a_j \in \mathbb{A}} \frac{1}{Q} \Pr(\mathbf{y} \mid x[c] = a_j, \mathbf{H}) \quad (32)$$

$$\approx \arg \max_{a_j \in \mathbb{A}} \prod_{d \in \mathcal{J}_c} \Pr(y[d] \mid x[c] = a_j, \mathbf{H}). \quad (33)$$

With channel estimation, how to decode?

-

$$\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{w},$$

Message passings from observation nodes $y[d]$ to variable nodes $x[c], c \in \mathcal{I}(d)$: The mean $\mu_{d,c}^{(i)}$ and variance $(\sigma_{d,c}^{(i)})^2$ of the interference, approximately modeled as a Gaussian random variable $\zeta_{d,c}^{(i)}$ defined as:

$$y[d] = x[c]H[d,c] + \underbrace{\sum_{e \in \mathcal{I}(d), e \neq c} x[e]H[d,e]}_{\zeta_{d,c}^{(i)}} + z[d], \quad (34)$$

can be computed as:

$$\mu_{d,c}^{(i)} = \sum_{e \in \mathcal{I}(d), e \neq c} \sum_{j=1}^Q p_{e,d}^{(i-1)}(a_j) a_j H[d,e], \quad (35)$$

and

$$(\sigma_{d,c}^{(i)})^2 = \sum_{e \in \mathcal{I}(d), e \neq c} \left(\sum_{j=1}^Q p_{e,d}^{(i-1)}(a_j) |a_j|^2 |H[d,e]|^2 - \left| \sum_{j=1}^Q p_{e,d}^{(i-1)}(a_j) a_j H[d,e] \right|^2 \right) + \sigma^2. \quad (36)$$

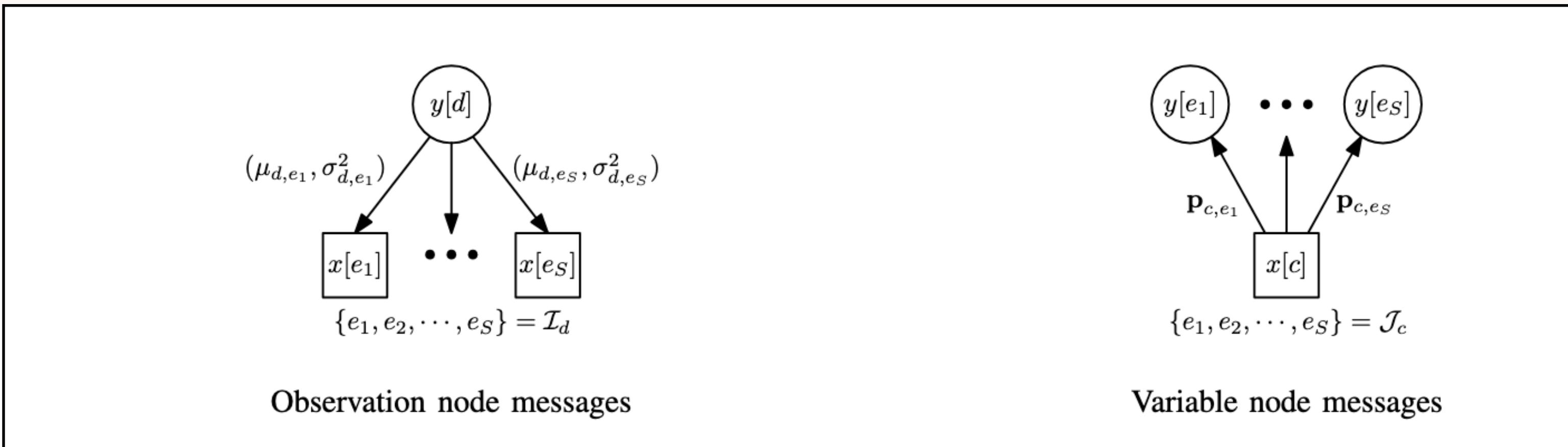
Message passings from variable nodes $x[c]$ to observation nodes $y[d], d \in \mathcal{J}(c)$: The pmf vector $\mathbf{p}_{c,d}^{(i)}$ can be updated as:

$$p_{c,d}^{(i)}(a_j) = \Delta \cdot \tilde{p}_{c,d}^{(i)}(a_j) + (1 - \Delta) \cdot p_{c,d}^{(i-1)}(a_j), a_j \in \mathbb{A} \quad (37)$$

where $\Delta \in (0, 1]$ is the *damping factor* used to improve the performance by controlling the convergence speed [18], and

$$\begin{aligned} \tilde{p}_{c,d}^{(i)}(a_j) &\propto \prod_{e \in \mathcal{J}(c), e \neq d} \Pr(y[e] \mid x[c] = a_j, \mathbf{H}) \\ &= \prod_{e \in \mathcal{J}(c), e \neq d} \frac{\xi^{(i)}(e, c, j)}{\sum_{k=1}^Q \xi^{(i)}(e, c, k)}, \end{aligned} \quad (38)$$

$$\text{where } \xi^{(i)}(e, c, k) = \exp \left(\frac{-|y[e] - \mu_{e,c}^{(i)} - H_{e,c} a_k|^2}{(\sigma_{e,c}^{(i)})^2} \right).$$



The message passing algorithm (copied from paper)

Algorithm 1 MP algorithm for OTFS symbol detection

Input: Received signal \mathbf{y} , channel matrix \mathbf{H} .

Initialization: pmf $\mathbf{p}_{c,d}^{(0)} = 1/Q, c = 1, \dots, NM, d \in \mathcal{J}(c)$, iteration count $i = 1$.

repeat

- Observation nodes $y[d]$ compute the means $\mu_{d,c}^{(i)}$ and variances $(\sigma_{d,c}^{(i)})^2$ of Gaussian random variables $\zeta_{d,c}^{(i)}$ using $\mathbf{p}_{c,d}^{(i-1)}$ and pass them to variable nodes $x[c], c \in \mathcal{I}(d)$.
- Variable nodes $x[c]$ update $\mathbf{p}_{c,d}^{(i)}$ using $\mu_{d,c}^{(i)}, (\sigma_{d,c}^{(i)})^2$, and $\mathbf{p}_{c,d}^{(i-1)}$ and pass them to observation nodes $y[d], d \in \mathcal{J}(c)$.
- Compute convergence indicator $\eta^{(i)}$.
- Update the decision on the transmitted symbols $\hat{x}[c], c = 1, \dots, NM$ if needed.
- $i = i + 1$

until *Stopping criteria*;

Output: The decision on transmitted symbols $\hat{x}[c]$.

With channel estimation, how to decode?

- Suppose one have estimated the (delay-Doppler domain) channel matrix $\mathbf{H} \in \mathbb{C}^{NM \times NM}$, then we are facing (equivalently) a linear system with NM equations:

$$\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{w},$$

- Where where $\mathbf{y}, \mathbf{x}, \mathbf{w}$ are $\mathbf{x}[k, l], \mathbf{y}[k, l], \mathbf{w}[k, l]$ in vectorized form, $k \in [N], l \in [M]$. \mathbf{H} is treated given.
-

- Easier ways to do decoding?
 - —MMSE! (*We use this in our MatLab code demo later*)

$$\hat{\mathbf{x}} \triangleq (\mathbf{H}^\dagger \mathbf{H} + \lambda \mathbf{I})^{-1} \mathbf{H}^\dagger \mathbf{y}$$

- Note that: this involves a matrix inversion of a $NM \times NM$ matrix. Whose (in general) complexity is $\mathcal{O}(M^3 N^3)$. Usually, when N, M are not too small. This will be huge...
- Side note: In fact, maybe we can leverage the property of \mathbf{H} being sparse to accelerate it. But the result is shown not good as MP.

What if there is no Doppler?

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OTFS Performance on Static Multipath Channels

P. Raviteja^{ID}, Emanuele Viterbo^{ID}, and Yi Hong^{ID}

- “..... In this letter, we **show that, in static multipath channels, the system structure of OTFS is equivalent to the asymmetric orthogonal frequency division multiplexing (A-OFDM, an OFDM-like scheme with unequal subcarrier spacing)**, a scheme ... bridging between cyclic prefix single carrier (CPSC) and traditional OFDM.

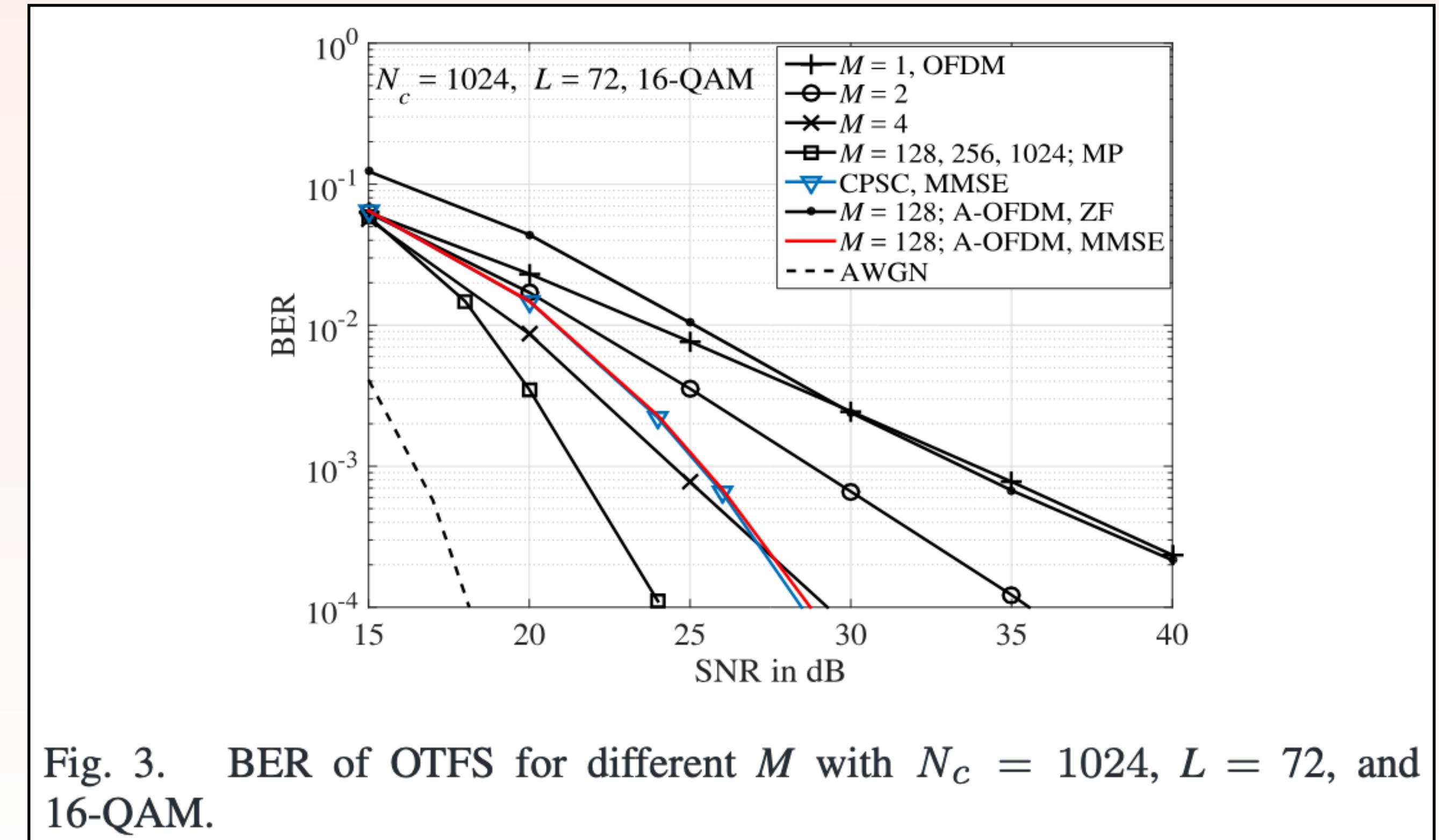


Fig. 3. BER of OTFS for different M with $N_c = 1024$, $L = 72$, and 16-QAM.

Matlab Code Demo

- Try some different SNRs and see the results.
- $N = 16, M = 64$. (The DD domain matrix is 16×64)
- Will output a number, indicating out of the $16 \times 64 = 1024$ data, how many are incorrect.
- 4QAM
- About ≈ 9 taps.
- Rayleigh fading + AWGN
- **Assuming channel matrix is given**

Other concepts related to OTFS, but not mentioned

- Quasi-periodicity
- Pulsone: (the waveform) of OTFS
- Zak transform: a useful tool in simplifying the modulation / demodulation of OTFS
- ...

OTFS community

- It was first presented in WCNC 2017, by Ronny Hadani.
- Since then, many (≥ 1000) works have been published.
- There is an official small community called “OTFS special interest group”. Members are experts and pioneers in OTFS. They host online webinar on a monthly basis.

Ronny Hadani's scholar profile shows four publications:

TITLE	CITED BY	YEAR
Orthogonal time frequency space modulation	1338	2017
Orthogonal time-frequency space modulation: A promising next-generation waveform	458	2021
OTFS methods of data channel characterization and uses thereof	303	2017
OTFS: A new generation of modulation addressing the challenges of 5G	301	2018



- We are thrilled to announce 2024's OTFS webinar series, organized monthly since March 2024, given by world-renowned experts in this area :
 - March 2024, **Summary of OTFS and its new developments**, given by Prof. Emanuele Viterbo and Prof. Yi Hong
 - April-July 2024, Talks on **Radar sensing with OTFS 2.0**, given by Prof. Saif Khan Mohammed, Prof. Ronny Hadani, Prof. A. Chockalingam, and Prof. Robert Calderbank
 - May 2024, **Orthogonal delay-Doppler multiplexing (ODDM)**, given by Prof. Jinhong Yuan and Prof. Hai Lin
 - June 2024, **Practical pulse shaping for DD communications based on the Zak transform**, given by Dr. Shuangyang Li
 - July 2024, **Recent results on delay-Doppler multiplexing**, given by Prof. Arman Farhang
 - August 2024, **A new pilot-free ISAC waveform**, given by Dr. Peter Jung and Dr. Philipp Walk
 - August 2024, **ODDM waveform**, given by Prof. Jinhong Yuan

References

- Yi Hong et al., “Delay-Doppler Communications: Principles and Applications”, Elsevier, 2022 [\[Very comprehensive monograph on DD communication and OTFS\]](#)
- Saif Khan Mohammed, , “OTFS - A Mathematical Foundation for Communication and Radar Sensing in the Delay-Doppler Domain”, IEEE, 2022.
- Emanuele Viterbo, “An introduction to Orthogonal Time Frequency Space (OTFS) modulation for high mobility communications”, 2019. SIG on OTFS. [\[slides\]](#)
- Emanuele Viterbo, “OTFS Detection and Channel Estimation: a Low-Complexity Practical Solution”, 2021. SIG on OTFS. [\[slides\]](#)
- P. Raviteja, Khoa T. Phan, Yi Hong, and Emanuele Viterbo, “Interference Cancellation and Iterative Detection for Orthogonal Time Frequency Space Modulation” [\[The paper talked about message passing decoding of OTFS\]](#)
- Brian Olson, Steven Shaw, Chengzhi Shi, Christophe Pierre, Robert Parker, “Circulant Matrices and Their Application to Vibration Analysis”

OTFS: channel matrix representation

- Received signal in vector form in time domain (assuming noiseless)

$$\mathbf{r} = \mathbf{G}\mathbf{s}$$

- \mathbf{G} is an $MN \times MN$ matrix of the following form

$$\mathbf{G} = \sum_{i=1}^P h_i' \boldsymbol{\Pi}^{l_i} \boldsymbol{\Delta}^{(k_i)},$$

where, $\boldsymbol{\Pi}$ is the permutation matrix (forward cyclic shift), and $\boldsymbol{\Delta}^{(k_i)}$ is the diagonal matrix

$$\underbrace{\boldsymbol{\Pi} = \begin{bmatrix} 0 & \dots & 0 & 1 \\ 1 & \ddots & 0 & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & \dots & 1 & 0 \end{bmatrix}}_{MN \times MN}, \quad \boldsymbol{\Delta}^{(k_i)} = \underbrace{\begin{bmatrix} e^{\frac{j2\pi k_i(0)}{MN}} & 0 & \dots & 0 \\ 0 & e^{\frac{j2\pi k_i(1)}{MN}} & \dots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \dots & e^{\frac{j2\pi k_i(MN-1)}{MN}} \end{bmatrix}}_{\text{Doppler}}$$

Delay (similar to OFDM)