

Capacity Analysis of Coprime Communication

Qiong Wu

University of Texas at Arlington
Wireless Comm. and Networking Lab

qiong.wu@mavs.uta.edu

December 13, 2013

Overview

- 1 Introduction
- 2 Preliminaries
- 3 System Modeling
- 4 Theorems of Capacity of Coprime Filter-banks
- 5 Conclusions

Introduction

The study of capacity of analog Gaussian channels and capacity-achieving transmission strategies was pioneered by Shannon, whose work focused on capacity of channels sampled at or above twice the channel bandwidth.

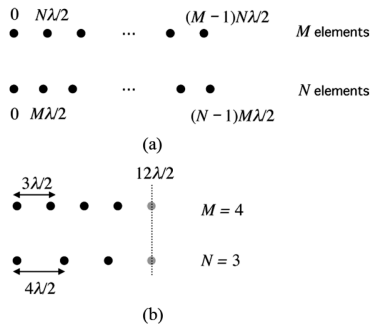
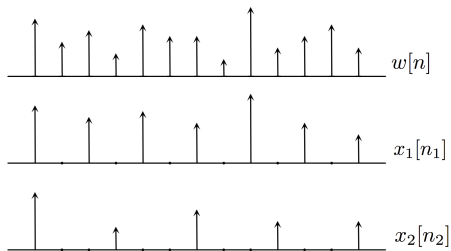
However, in practice, the Nyquist rate may be **excessive** for perfect reconstruction of signals that the transmission channel possess certain structures known a priori. On the other hand, the **hardware** and **power limitations** may preclude sampling at the Nyquist rate for a wideband communication system.

This motivates the exploration of the effects of sub-Nyquist sampling upon the capacity of an analog Gaussian channel, and the capacity limits that result from general sampling methods.

Related works

- When the channel or signal structure is unknown, the blind sub-Nyquist sampling approaches have been proposed to exploit the structure of various classes of input signals based on sampling with modulation and filter banks [Mishali, 2010].
- At the transmitter side, although MRC maximizes the combiner SNR for a MISO channel, it is suboptimal for the joint optimization problem compared with selection combining [Goldsmith, 2005].
- The capacity with sampling rate under filter-banks with/without modulation are not monotonously increasing, which indicates that more sophisticated sampling techniques are necessary to maximize the capacity [Chen, 2013].

Preliminaries: Revisit Coprime



Theorem 1: Channel capacity

[4, Th. 8.5.1] For the channel with a power constraint P and noise PSD $N(f)$, assume that $|H(f)|^2/N(f)$ is bounded and integrable, and that either $\int_{-\infty}^{+\infty} N(f)df < +\infty$ or that $N(f)$ is white. Then the channel capacity is given by

$$\mathbb{C} = \frac{1}{2} \int_{-\infty}^{+\infty} \log^+ \left(\nu \frac{|H(f)|^2}{N(f)} \right) df, \quad (1)$$

where ν is the water-filling for additive Gaussian channel and satisfying

$$\int_{-\infty}^{+\infty} \left[\nu - \frac{N(f)}{|H(f)|^2} \right]^+ df = P. \quad (2)$$

Theorem 2: Landau Sampling Rate

[5, Th. 1 and Lemma 4] Given a fixed set S of N -frequency bands whose total finite length, including both the negative and positive frequencies, denoted by $m(S)$, the space $\mathcal{B}(S)$ of all signals of finite energy with frequencies contained only in S , and Λ is a sampling set from $\mathcal{B}(S)$, All signals $f(t) \in \mathcal{B}(S)$ can be uniquely determined by their samples $\{f(t_n) | t_n \in \Lambda\}$ only if the lower Beurling density $D^-(\Lambda)$ satisfying

$$\begin{aligned} D^-(\Lambda) &= \lim_{r \rightarrow \infty} \inf \frac{n^-(r)}{r} \\ &\geq \lim_{r \rightarrow \infty} \left[\frac{m(S)}{2\pi} - \frac{A \log^+(r) - B}{r} \right] \end{aligned} \quad (3)$$

with constant A and B relevant with the choices of S and Λ and independent of the sampling interval r .

Basics of Toeplitz Matrix

$$T_n = \begin{bmatrix} t_0 & t_{-1} & t_{-2} & \dots & t_{-(n-1)} \\ t_1 & t_0 & t_{-1} & \dots & t_{-(n-2)} \\ t_2 & t_1 & t_0 & \dots & t_{-(n-3)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ t_{n-1} & t_{n-2} & t_{n-3} & \dots & t_0 \end{bmatrix} \quad (4)$$

- Toeplitz matrices are also closely connected with Fourier series, and has the property of commute asymptotically.
- A Toeplitz matrix can be briefly described as $T_n = [t_{i-j}; i, j = 0, 1, \dots, n-1]$, which is uniquely defined by the sequence $\{t_k\}$.

Basics of Toeplitz Matrix

The $n \times n$ Toeplitz matrix generated by Fourier coefficients is

$$T_n(f) = \left[\int_0^{2\pi} f(\lambda) e^{j(i-k)\lambda} \frac{d\lambda}{2\pi}; \quad k, i = 0, 1, \dots, n-1 \right]. \quad (5)$$

Besides, let U_n be the circulant matrix of T_n by filling in the upper right and lower left corners with the appropriate entries. In particular, U_n consists of cyclic shifts of $[c_0^{(n)}, \dots, c_{n-1}^{(n)}]$ where

$$c_k^{(n)} = \begin{cases} t_{-k} & k = 0, 1, \dots, m \\ t_{n-k} & k = n-m, \dots, n-1. \\ 0 & \text{otherwise} \end{cases} \quad (6)$$

Similarly, the Fourier coefficients of the circulant matrix $U_n(f)$ is also well defined

$$U_n(f) = \left[\frac{1}{n} \sum_{j=0}^{n-1} f\left(\frac{2\pi j}{n}\right) e^{(\frac{2\pi jk}{n})}; i, k = 0, 1, \dots, n-1 \right]. \quad (7)$$

Definition: Asymptotic Equivalence of Toeplitz Matrix

[6] Two $n \times n$ matrices $\{A_n\}$ and $\{B_n\}$ are said to be asymptotically equivalent and be abbreviated as $A_n \sim B_n$ if

- A_n and B_n are uniformly bounded for any integer n and a constant Ψ independent of n :

$$\|A_n\|_2, \|B_n\|_2 \leq \Psi < +\infty \quad (8)$$

- The determinant of the difference between A_n and B_n is approximately zero as $n \rightarrow \infty$.

$$\lim_{n \rightarrow \infty} |A_n - B_n| = 0 \quad (9)$$

Lemma 1: Eigenvalues of Toeplitz Matrices

[6, Lemma 1] Suppose $A_n \sim B_n$ with eigenvalues $\{\alpha_{n,k}\}$ and $\{\beta_{n,k}\}$, respectively. Let $g(x)$ be an arbitrary continuous function. Then,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} (\alpha_{n,k} - \beta_{n,k}) = 0 \quad (10)$$

and hence if either limit exists individually,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} \alpha_{n,k} = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} \beta_{n,k} \quad (11)$$

Lemma 2: Operations of Toeplitz Matrices

[6, Lemma 7 and 10]

- Suppose a banded Toeplitz matrix $T_n = \{t_{i-j}; i, j = 0, 1, \dots, n-1\}$, which satisfies that $\{t_k\}$ is absolutely summable, and the Fourier series $f(\lambda)$ related to T_n is positive and T_n is Hermitian. Then we have

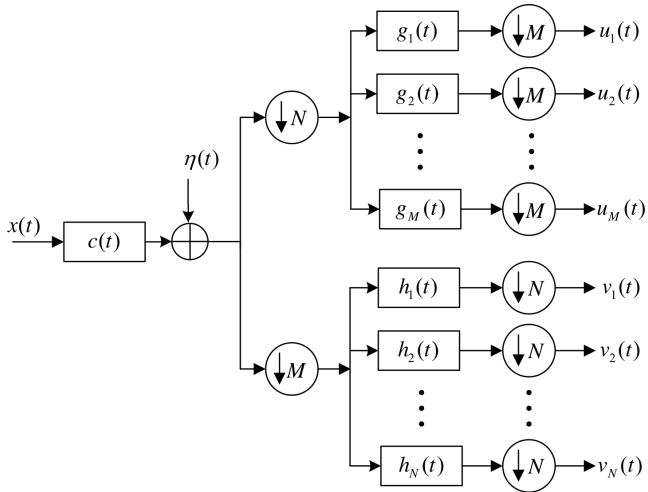
$$T_n(f) \sim U_n(f) \quad (12)$$

If $f(\lambda)$ is real and there exists a constant $\varepsilon > 0$ such that $\inf f \geq \varepsilon$, then

$$T_n(f)^{-1} \sim U_n(f)^{-1} = U_n(1/f) \sim T_n(1/f) \quad (13)$$

- Suppose $A_n \sim B_n$ and $C_n \sim D_n$, then $A_n C_n \sim B_n D_n$.

System Modeling



Matrices of Impulse Responses

$$\mathbf{G}_m = [\dots G_{m,-1} \ G_{m,0} \ G_{m,1} \ \dots] \quad 0 \leq m \leq M \quad (14)$$

$$G_{m,0} = [g_m^0, \underbrace{0, \dots, 0}_{N-1}, g_m^1, 0, \dots, 0, g_m^{\frac{L-1}{M}}]^T \quad (15)$$

$$\mathbf{G}_m = \begin{bmatrix} \dots & g_m^0 & 0 & \dots & 0 & g_m^{-N} & \dots \\ \dots & 0 & g_m^0 & \dots & 0 & 0 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \dots & 0 & 0 & \dots & g_m^0 & 0 & \dots \\ \dots & g_m^N & 0 & \dots & 0 & g_m^0 & \dots \\ \dots & 0 & g_m^N & \dots & 0 & 0 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \dots & g_m^{\frac{L-1}{M}} & 0 & \dots & 0 & g_m^{\frac{L-N-1}{M}} & \dots \end{bmatrix}. \quad (16)$$

Equivalent Filter Impulse Responses

We also define the *virtual* channel impulse response $\tilde{c}(t)$ in terms of *equivalent* filter impulse response $s(t)$. For the i th branch, it is

$$\tilde{c}_i(t) = c(t) * s_i(t) \quad 1 \leq i \leq MN$$

$$\tilde{\mathbf{c}}_i^l = \left[\tilde{c}_i(\alpha \tilde{T}_s), \tilde{c}_i(\alpha \tilde{T}_s - \xi), \dots, \tilde{c}_i(\alpha \tilde{T}_s - (\beta - 1)\xi) \right],$$

where $0 \leq \alpha \leq L$. Specifically, the matrix representations are

$$\tilde{\mathbf{C}}_i = \begin{bmatrix} \tilde{\mathbf{c}}_i^0 & \tilde{\mathbf{c}}_i^{-1} & \dots & \tilde{\mathbf{c}}_i^{-L+1} \\ \tilde{\mathbf{c}}_i^1 & \tilde{\mathbf{c}}_i^0 & \dots & \tilde{\mathbf{c}}_i^{-L+2} \\ \vdots & \vdots & \ddots & \vdots \\ \tilde{\mathbf{c}}_i^{L-1} & \tilde{\mathbf{c}}_i^{L-2} & \dots & \tilde{\mathbf{c}}_i^0 \end{bmatrix} \quad (17)$$

$$\mathbf{S}_i = \begin{bmatrix} s_i^{L-1} & \dots & s_i^0 & s_i^{-1} & \dots & s_i^{-L+1} \\ s_i^{-L+1} & \dots & s_i^1 & s_i^0 & \dots & s_i^{-L+2} \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ s_i^{-1} & \dots & s_i^{L-1} & s_i^{L-2} & \dots & s_i^0 \end{bmatrix} \quad (18)$$

Equivalent Filter Impulse Responses

Considering the virtual filter impulse responses \mathbf{S} in (18) and any $1 \leq i \leq j \leq MN$, we have

$$(\mathbf{S}_u \mathbf{S}_v^*)_{i,j} = (\mathbf{S}_u \mathbf{S}_v^*)_{i,j}^* = \sum_{t=-\infty}^{\infty} s_u^{j-i+t} (s_v^t)^* \quad (19)$$

Hence, the Hermitian matrix $\tilde{\mathbf{S}} := \mathbf{S}_u \mathbf{S}_v^*$ is still Toeplitz, which is used to relate the matrix of interest in the capacity $(\mathbf{G}_M \mathbf{H}_N^*)^{-1/2}$ with the asymptotically equivalent circulant matrix \mathbf{U} defined in (7), which is able to preserve the Toeplitz property as being taken the inverse square root $(\mathbf{U})^{-1/2}$.

Theorem 3: Asymptotic Equivalence for Coprime Filter-banks

If there exists some constant $\varepsilon > 0$ such that for all $f \in \left[-\frac{f_s}{2MN}, \frac{f_s}{2MN}\right]$,

$$\sum_{l \in \mathbb{Z}} |G(f - lf_s)H(f - lf_s)| \geq \varepsilon > 0 \quad (20)$$

holds, then $(\mathbf{U})^{-1/2} \sim (\mathbf{G}_M \mathbf{H}_N^*)^{-1/2}$.

Theorem 3: Capacity of Coprime Filter-banks

Assume that $c(t)$ and $s_i(t)$ ($1 \leq i \leq MN$) are all continuous, bounded and absolutely Riemann integrable, and $c_\eta(t) = \mathcal{F}^{-1} \left(\frac{C(f)}{\sqrt{S_\eta(f)}} \right)$ satisfies $c_\eta(t) = o(t^{-\varepsilon})$ for some constant $\varepsilon > 1$, and that $\mathbf{F}_s(f)$ is right-invertible for every f . Define $\tilde{\mathbf{F}}_s \triangleq (\mathbf{F}_s \mathbf{F}_s^*)^{-\frac{1}{2}} \mathbf{F}_s$. The capacity $\mathbb{C}(f_s)$ of the sampled channel with a power constraint P is given as

$$\mathbb{C}(f_s) = \frac{1}{2} \int_{-\frac{f_s}{2MN}}^{\frac{f_s}{2MN}} \sum_{i=1}^{nMN} \log^+ \left(\nu \lambda_i \left(\tilde{\mathbf{F}}_s \mathbf{F}_c \mathbf{F}_c^* \tilde{\mathbf{F}}_s^* \right) \right) df \quad (21)$$

where the ν is satisfying

$$\int_{-\frac{f_s}{2MN}}^{\frac{f_s}{2MN}} \sum_{i=1}^{nMN} \left(\nu - \frac{1}{\lambda_i \left(\tilde{\mathbf{F}}_s \mathbf{F}_c \mathbf{F}_c^* \tilde{\mathbf{F}}_s^* \right)} \right)^+ df = P \quad (22)$$

Corollary: Existence of the Maximum Capacity

For each aliased set $\{f - \frac{if_s}{MN} | i \in \mathbb{Z}\}$ and each k ($1 \leq k \leq MN$), there exists an integer l such that $\frac{|H(f - \frac{lf_s}{MN})|^2}{S_\eta(f - \frac{lf_s}{MN})}$ is equal to the k th largest element in $\left\{ \frac{|H(f - \frac{lf_s}{MN})|^2}{S_\eta(f - \frac{lf_s}{MN})} | i \in \mathbb{Z} \right\}$. The maximum capacity is

$$\mathbb{C}(f_s) = \frac{1}{2} \int_{-\frac{f_s}{2MN}}^{\frac{f_s}{2MN}} \sum_{i=1}^{nMN} \log^+ (\nu \lambda_i (\mathbf{F}_c \mathbf{F}_c^*)) df \quad (23)$$

under the conditions that...

Corollary: Existence of the Maximum Capacity (cont.)

the frequency response of the k th filter of the filter bank is given by

$$S_i \left(f - \frac{lf_s}{MN} \right) = \begin{cases} 1, & \frac{|H(f - \frac{lf_s}{MN})|^2}{S_\eta(f - \frac{lf_s}{MN})} = \lambda_i (\mathbf{F}_c \mathbf{F}_c^*) \\ 0, & \text{otherwise,} \end{cases} \quad (24)$$

and the corresponding water filling scheme for ν is

$$\int_{-\frac{f_s}{2MN}}^{\frac{f_s}{2MN}} \sum_{i=1}^{nMN} \left(\nu - \frac{1}{\lambda_i (\mathbf{F}_c \mathbf{F}_c^*)} \right)^+ df = P \quad (25)$$

References



M. Mishali and Y. C. Eldar, "From theory to practice: sub-Nyquist sampling of sparse wideband analog signals," *IEEE J. Sel. Topics Signal Process.*, vol. 4, no. 2, pp. 375-391, Apr. 2010.



A. J. Goldsmith, *Wireless Communications*. New York, USA: Cambridge Univ. Press, 2005.



Y. Chen, Y. C. Eldar, and A. J. Goldsmith, "Shannon meets Nyquist: capacity of sampled Gaussian channels," *IEEE Transactions on Information Theory*, vol. 59, no. 8, pp. 4889-4914, Aug. 2013.



R. G. Gallager, *Information theory and reliable communication*, New York, Wiley, 1968.



H. Landau, "Necessary density conditions for sampling and interpolation of certain entire functions," *Acta Mathematica*, vol. 117, pp. 37-52, 1967.



R. Gray, "Toeplitz and circulant matrices: A review", *Foundations and Trends in Communications and Information Theory*, vol. 2, no. 3, pp. 155-239, Now Publishers, Delft, The Netherlands, 2006.

Thanks!