

Background

Matrix Products for Column Vectors in \mathbb{R}^d

- ▶ The **dot product** of \bar{x}, \bar{y} in \mathbb{R}^d is $\bar{x}^T \bar{y} \in \mathbb{R}$. The length of \bar{x} satisfies $\|\bar{x}\| = \sqrt{\bar{x}^T \bar{x}}$.
- ▶ The **tensor product** is the $d \times d$ matrix $\bar{x} \bar{y}^T$. The tensor product $\bar{x} \bar{x}^T$ is **symmetric**, and **positive semi-definite** since

$$\bar{y}^T (\bar{x} \bar{x}^T) \bar{y} = (\bar{x}^T \bar{y})^2 \geq 0.$$

Frame Theory

- ▶ A **FUNTF** is a **Finite Unit-Norm Tight Frame**.
- ▶ A real **finite frame** is a collection of vectors $\{\bar{x}_i\}_{i=1}^N$ in \mathbb{R}^d for which there exists constants (the **frame bounds**) $0 < A \leq B$ satisfying

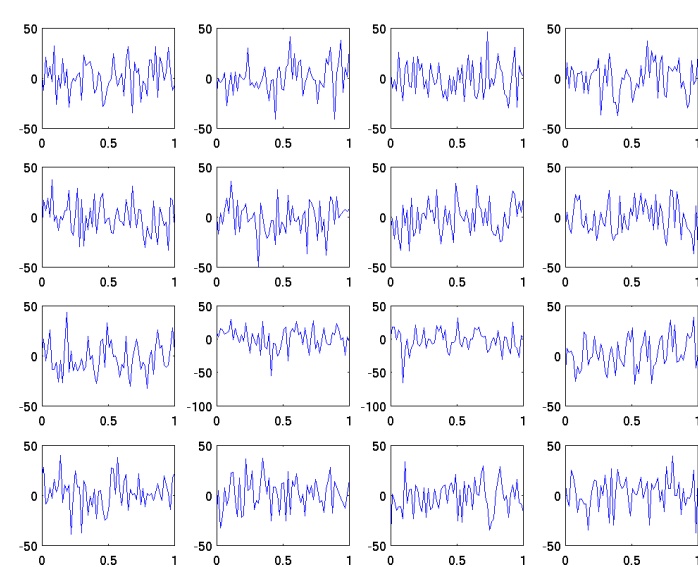
$$A \|\bar{y}\|^2 \leq \sum_{i=1}^N \bar{y}^T (\bar{x}_i \bar{x}_i^T) \bar{y} \leq B \|\bar{y}\|^2 \text{ for all } \bar{y} \in \mathbb{R}^d.$$

- ▶ If $A = B$, then the frame is called **tight**. If $\|\bar{x}_i\| = 1$ for all $i = 1, \dots, N$, then the frame is called **unit-norm**.
- ▶ The **frame operator** is $S = \sum_{i=1}^N \bar{x}_i \bar{x}_i^T$, so $A \|\bar{y}\|^2 \leq \bar{y}^T S \bar{y} \leq B \|\bar{y}\|^2$ for all $y \in \mathbb{R}^d$.
- ▶ The frame operator of a FUNTF is a multiple of the $d \times d$ identity matrix: $\frac{N}{d} I_{d \times d}$. It is also positive definite, so we write $0 < S$.

Why Design FUNTFs?

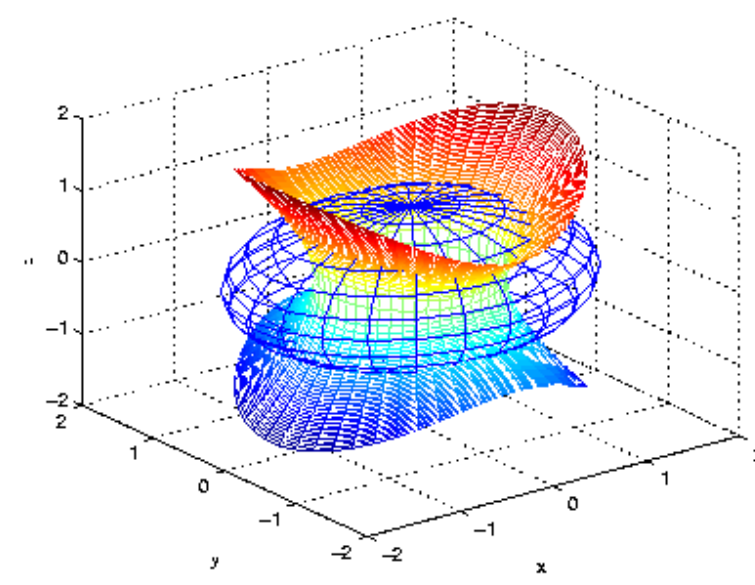
Synchronous CDMA

- ▶ **C**ode **D**ivision **M**ultiple **A**ccess
- ▶ Used for cell phones
- ▶ Need Optimal Codes

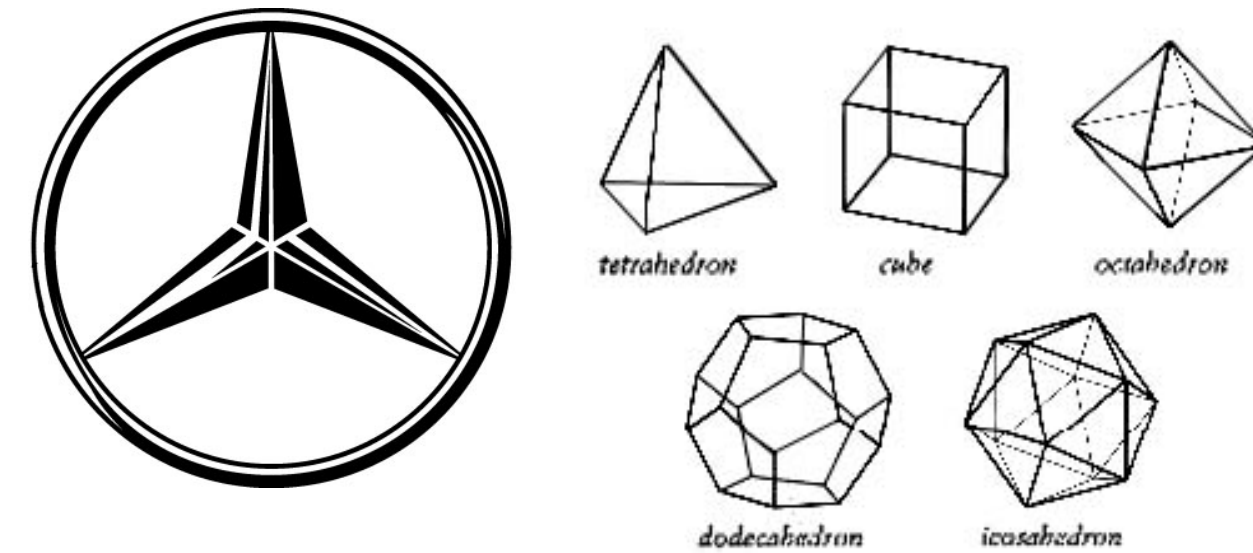


Beautiful Mathematics

- ▶ Kadison-Singer conjecture
- ▶ Algebraic geometry



Examples of FUNTFs



The Constraints for Frame Design

Suppose $A \in M_d(\mathbb{R})$ is symmetric and positive definite ($0 < A$) and let $\bar{x} \in \mathbb{R}^d$. Then

$$0 \leq A - \bar{x} \bar{x}^T \iff \bar{x}^T A^{-1} \bar{x} \leq 1$$

and

$$\text{rank}(A - \bar{x} \bar{x}^T) < \text{rank}(A) \iff \bar{x}^T A^{-1} \bar{x} = 1$$

Proof.

Note that $A - \bar{x} \bar{x}^T$ has at most one non-negative eigenvalue by the interlacing inequalities for eigenvalues. Now, the determinant of $A - \bar{x} \bar{x}^T$ is the product of the eigenvalues, so we infer that this difference will be positive semi-definite if and only if the determinant is non-negative.

Since A is positive definite, it is invertible, and we may use the **Sherman-Morrison determinant formula**:

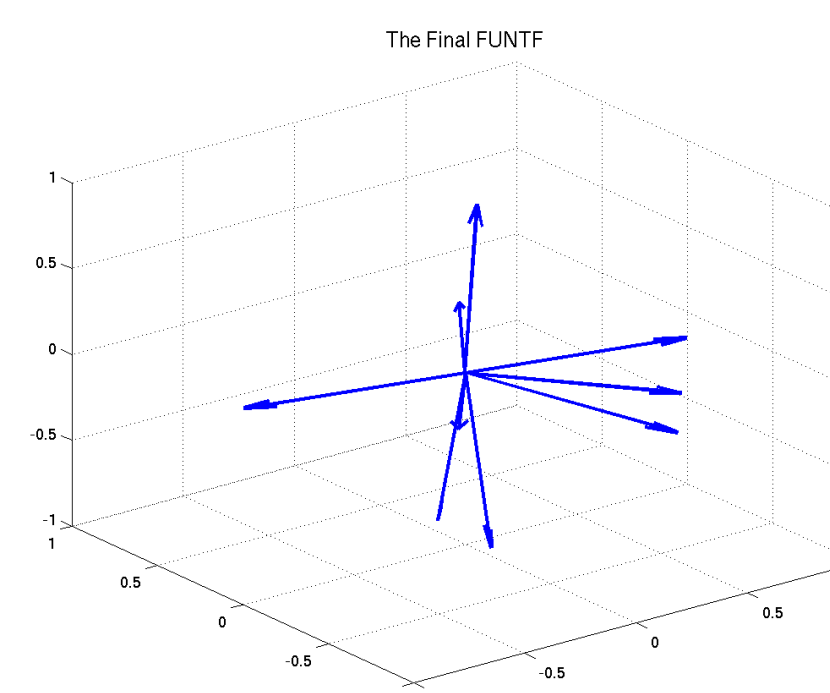
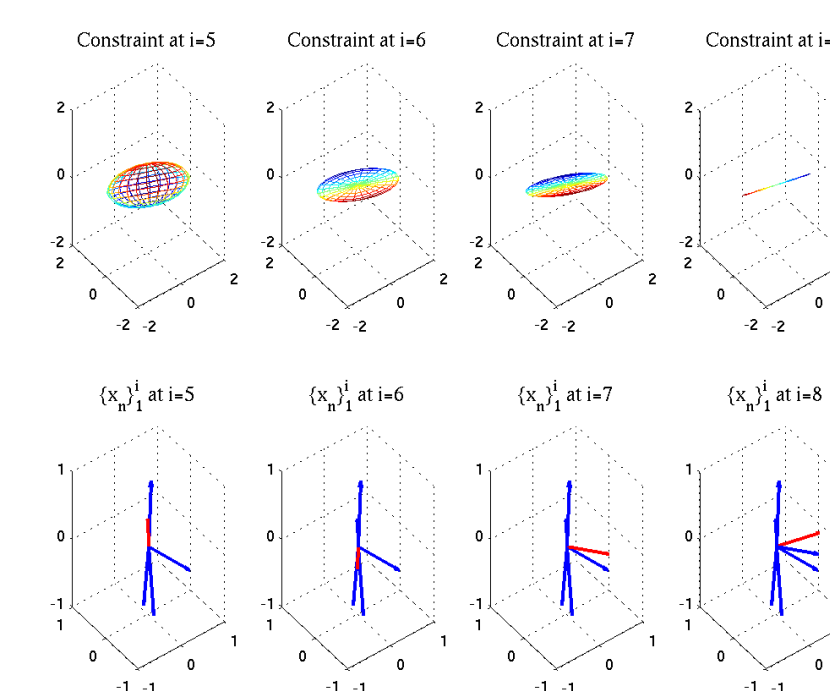
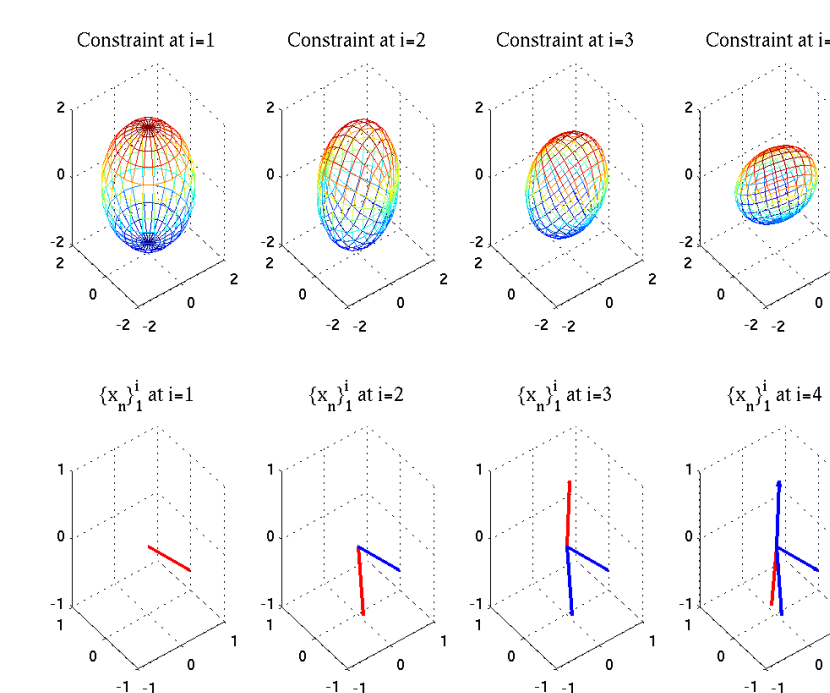
$$\det(A - \bar{x} \bar{x}^T) = \det(A) (1 - \bar{x}^T A^{-1} \bar{x}).$$

But $\det(A) > 0$ since A is positive definite, so the left term is non-negative if and only if $0 \leq 1 - \bar{x}^T A^{-1} \bar{x}$. Thus, $A - \bar{x} \bar{x}^T$ is positive semi-definite if and only if $\bar{x}^T A^{-1} \bar{x} \leq 1$. The rank inequality follows from similar considerations. \square

FUNTF Design Algorithm

- (0) Input N, d . Set $S = (N/d) I_{d \times d}$.
- (1) For $i = \{1, \dots, N-d\}$
 - ▶ Choose \bar{x}_i : $\bar{x}_i^T S^{-1} \bar{x}_i < 1$.
 - ▶ $S \leftarrow S - \bar{x}_i \bar{x}_i^T$.
- (2) For $i = \{N-d+1, \dots, N\}$
 - ▶ Set $\tilde{S}^{-1} \bar{y} = \bar{v}$: $\bar{v} \in \text{Image}(S)$, $S \bar{v} = \bar{y}$.
 - ▶ Choose \bar{x}_i : $\bar{x}_i \in \text{Image}(S)$, $\bar{x}_i^T \tilde{S}^{-1} \bar{x}_i = 1$.
 - ▶ $S \leftarrow S - \bar{x}_i \bar{x}_i^T$.

Example: $N=9, d=3$



The Sherman-Morrison Determinant Formula

$$\det(A - \bar{x} \bar{x}^T) = \det(A) \cdot (1 - \bar{x}^T A^{-1} \bar{x})$$

Proof.

Note that A must be invertible to apply this formula. Consider the case $d = 2$ (this proof is easily adapted to general d). Setting $A = [\bar{a}_1 \ \bar{a}_2]$ and $\bar{x} = [x_1 \ x_2]^T$, we have that

$$\begin{aligned} \det(A - \bar{x} \bar{x}^T) &= \det([\bar{a}_1 - x_1 \bar{x} \ \bar{a}_2 - x_2 \bar{x}]) \\ &= \det([\bar{a}_1 \ \bar{a}_2]) - \det([x_1 \bar{x} \ \bar{a}_2]) \\ &\quad - \det([\bar{a}_1 \ x_2 \bar{x}]) + \det([x_1 \bar{x} \ x_2 \bar{x}]) \\ &= \det(A) - x_1 \det(A_1(\bar{x})) - x_2 \det(A_2(\bar{x})) + 0 \\ &= \det(A) - \det(A) x_1 (A^{-1} \bar{x})_1 - \det(A) x_2 (A^{-1} \bar{x})_2 \\ &= \det(A) - \det(A) (\bar{x}^T A^{-1} \bar{x}) \\ &= \det(A) (1 - \bar{x}^T A^{-1} \bar{x}). \end{aligned}$$

To get to line (1), we used the fact that the determinant is a multi-linear function of the columns. To get to line (2), we note that the determinant of a matrix with linearly dependent columns is zero. To get to line (3), we use Cramer's rule. \square

References

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