Maths for Signals and Systems Linear Algebra in Engineering

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Mathematics for Signals and Systems

In this set of lectures we will talk about two applications:

- Discrete Fourier Transforms
- An application of linear system theory: graphs and networks

The Discrete Fourier Transform (DFT) matrix

• The $n \times n$ Fourier matrix is defined as:

$$F_n = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & w & w^2 & \dots & w^{(n-1)} \\ 1 & w^2 & w^4 & \dots & w^{2(n-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & w^{(n-1)} & w^{2(n-1)} & \dots & w^{(n-1)(n-1)} \end{bmatrix}$$

- In this matrix we will number the first row and column with 0.
- We define $w = e^{-i\frac{2\pi}{n}}$. For w is preferable to use polar representation.
- $F_n(i,j) = w^{ij}$.
- We must stress out that it is better to use the notation w_n instead of w.
- I have, in general, avoided this notation to make things look simpler but occasionally I used it.

The Discrete Fourier Transform (DFT) matrix cont.

• The parameter $w = e^{-i\frac{2\pi}{n}}$ lies on the unit circle shown below. The case depicted below refers to n = 8 where the points $w^m = e^{-i\frac{2\pi m}{8}}$, m = 0, ..., 7 of the second row (row 1) of the Fourier matrix are shown.



• We must stress out that the Fourier matrix is totally constructed out of numbers of the form w_n^k .

The Discrete Fourier Transform (DFT) matrix for n=4

- The parameter $w_4 = e^{-i\frac{2\pi}{4}} = e^{-i\frac{\pi}{2}} = \cos\left(-\frac{\pi}{2}\right) + i\sin\left(-\frac{\pi}{2}\right) = -i.$
- The quantities inside Fourier matrix are 1, i, , i², i³, i⁴, i⁶, i⁹.
- , i²

$$F_4 = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -\mathbf{i} & (-\mathbf{i})^2 & (-\mathbf{i})^3 \\ 1 & (-\mathbf{i})^2 & (-\mathbf{i})^4 & (-\mathbf{i})^6 \\ 1 & (-\mathbf{i})^3 & (-\mathbf{i})^6 & (-\mathbf{i})^9 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -\mathbf{i} & (-\mathbf{i})^2 & (-\mathbf{i})^3 \\ 1 & (-\mathbf{i})^2 & (-\mathbf{i})^0 & (-\mathbf{i})^2 \\ 1 & (-\mathbf{i})^3 & (-\mathbf{i})^2 & (-\mathbf{i})^1 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -\mathbf{i} & -1 & \mathbf{i} \\ 1 & -1 & 1 & -1 \\ 1 & \mathbf{i} & -1 & -\mathbf{i} \end{bmatrix}$$

- The columns of this matrix are orthogonal.
- Remember that the inner product of 2 complex vectors is $y^{*^{T}}x = y^{H}x$.

The Discrete Fourier Transform (DFT) matrix for n=4 cont.

- I can show that the columns are orthogonal but they are not orthonormal.
- I can fix this by dividing the Fourier matrix with the length of the rows (columns). In the case of n = 4 it is 2. Therefore, I can write:

$$F_4 = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -\mathbf{i} & -1 & \mathbf{i} \\ 1 & -1 & 1 & -1 \\ 1 & \mathbf{i} & -1 & -\mathbf{i} \end{bmatrix}$$

• We can easily show that $F_4^{H}F_4 = I$.

The Fast Fourier Transform (FFT)

- It can be proven that there is a connection between F_{2n} and F_n .
- This is expected from the fact that $w_{2n}^2 = e^{-i\frac{4\pi}{2n}} = e^{-i\frac{2\pi}{n}} = w_n$. It can be shown that: diagonal matrix D_n $[F_{2n}] = \begin{bmatrix} I_n & D_n \\ I_n & -D_n \end{bmatrix} \begin{bmatrix} F_n & \mathbf{0}_n \\ \mathbf{0}_n & F_n \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & \cdots \\ 0 & 0 & 1 & 0 & 0 & 0 & \cdots \\ 0 & 1 & 0 & 0 & 0 & 1 & \cdots \\ 0 & 1 & 0 & 0 & 0 & \cdots \\ 0 & 0 & 1 & 0 & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix}$ permutation matrix P_{2n}
- When $[F_{2n}]$ is multiplied by a column vector in order to obtain the Fourier Transform of the signal, we require $(2n)^2$ multiplications.
- When $[F_{2n}]$ is decomposed as above, P_{2n} does not contribute to multiplications, $\begin{bmatrix} F_n & \mathbf{0}_n \\ \mathbf{0}_n & F_n \end{bmatrix}$ requires $2 \times (n)^2$ multiplications and $\begin{bmatrix} I_n & D_n \\ I_n & -D_n \end{bmatrix}$ requires n multiplications.
- In total $2 \times (n)^2 + n < (2n)^2$.



The Fast Fourier Transform (FFT) cont.

• In the previous analysis the matrix D_n is defined as:



• We start requiring $(2n)^2$ multiplications and manage to reduce them to $2 \times (n)^2 + n$ multiplications.

The Fast Fourier Transform (FFT) cont.

• The next step is to break the F_n down. We use the above idea recursively.

$$\begin{bmatrix} F_{2n} \end{bmatrix} = \begin{bmatrix} I_n & D_n \\ I_n & -D_n \end{bmatrix} \begin{bmatrix} F_n & \mathbf{0}_n \\ \mathbf{0}_n & F_n \end{bmatrix} P_{2n} = \\ \begin{bmatrix} I_n & D_n \\ I_{n/2} & -D_{n/2} \end{bmatrix} \begin{bmatrix} I_{n/2} & \mathbf{0}_{n/2} \\ \mathbf{0}_{n/2} & F_{n/2} \end{bmatrix} \begin{bmatrix} F_{n/2} & \mathbf{0}_{n/2} \\ \mathbf{0}_{n/2} & F_{n/2} \end{bmatrix} \begin{bmatrix} P_n & \mathbf{0}_n \\ \mathbf{0}_{n/2} & F_{n/2} \end{bmatrix} \begin{bmatrix} P_n & \mathbf{0}_n \\ \mathbf{0}_n & P_n \end{bmatrix} P_{2n}$$

- We started with $(2n)^2$ multiplications and manage to reduce them to $2 \times (n)^2 + n$ multiplication.
- Now the n^2 multiplications are reduced to $2 \times (n/2)^2 + n/2$ multiplications.

The Fast Fourier Transform (FFT) cont.

- We can carry on this recursive procedure until we reach 1×1 Fourier matrices.
- We will have a large number of matrices piling up.
- It can be proven that if we start with a matrix of size $n \times n$ the total number of multiplications is reduced to

$$\frac{1}{2}n\log_2(n)$$

- Consider $n = 1024 = 2^{10}$. In that case $n^2 > 1,000,000$.
- $\frac{1}{2}$ 1024log₂(1024) = 5 × 1024.
- We reduced the multiplications from 1024×1024 to 5×1024 , i.e., by a factor of 200.
- The Fast Fourier Transform is one of the most important algorithms in modern scientific computing.

Directed graphs and networks: The incidence matrix

- A graph is a mathematical model which consists of a set of nodes and edges denoted as: Graph={nodes, edges}
- Graphs are used in various applications.
- A graph can be represented by a matrix called **incidence matrix**.

$$A = \begin{bmatrix} -1 & 1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 1 \\ 0 & 0 & -1 & 1 \end{bmatrix}$$

- Each row corresponds to an edge. Row numbers are edge numbers.
- Each column corresponds to a node. Column numbers are node numbers.
- The element $A_{ij} = 1$ if an arrow points towards node *j* accross edge *i*.
- The element $A_{ij} = -1$ if an arrow leaves node *j* accross edge *i*.



Types of graphs

A graph where every pair of nodes is connected with an edge is a complete graph.
 It has the maximum number of edges m = ¹/₂n(n - 1) where m is the number of edges and n is the number of nodes.



• A graph without closed loops is a **tree**. It has the minimum number of edges m = n - 1.



Directed graphs and networks: The incidence matrix.

• The graph given previously is neither a complete graph nor a tree.



• Let us focus again on the **incidence matrix**.

	[-1	1	0	ך0
	0	-1	1	0
A =	-1	0	1	0
	-1	0	0	1
	L 0	0	-1	1

- Observe that Row 3 = Row 1 + Row 2.
 - We also observe that edges 1, 2 and 3 form a closed loop.
 - We can make the statement that closed loops correspond to dependent rows.
 - Independent rows come from trees.
 - rank(A) = 3. Independent rows are 1, 2, 4 or 1, 2, 5 or 1, 4, 5 or 2, 4, 5.
 - Furthermore, rank(A) = 3 tells us that after 3 edges we start forming loops.

Graphs and networks. Potential differences Ax

• Let us find the null space of the matrix that corresponds to the graph of interest:

$$Ax = \mathbf{0} \Rightarrow \begin{bmatrix} -1 & 1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 1 \\ 0 & 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} x_2 - x_1 \\ x_3 - x_2 \\ x_3 - x_1 \\ x_4 - x_1 \\ x_4 - x_3 \end{bmatrix} = \mathbf{0}$$

- In a real life circuit Ax is a vector of potential or voltage differences.
- If the graph represents and electronic circuit, the elements of vector *x* may represent potentials at nodes (e.g. voltages).
- $x_i x_j$ represents the difference in potential across certain edges.
- We see that a solution of the above system is $x = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}^T$.
- The null space is formed by vectors $c\begin{bmatrix}1 & 1 & 1\end{bmatrix}^T$ and $\dim(N(A)) = 1$.
- The solution to the above system is obtained subject to a scalar *c*.
- Since n = 4 and $\operatorname{and} \operatorname{dim}(N(A)) = 1$, we see again that $\operatorname{rank}(A) = 3$.
- By fixing the potential at node one to 0 we remove the first column and we solve for the remaining potentials. In Electrical Engineering this is translated as node 1 been "grounded".

Graphs and networks: Kirchoff's Current Law (KCL)

• Let us consider the equation

$$A^{T}y = 0 \Rightarrow \begin{bmatrix} -1 & 0 & -1 & -1 & 0 \\ 1 & -1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & -1 \\ 0 & 0 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} y_{1} \\ y_{2} \\ y_{3} \\ y_{4} \\ y_{5} \end{bmatrix} = \mathbf{0}$$

- The vector *y* represents currents across the edges (or it could represent a force).
- The equation A^Ty = 0 represents Kirchoff's Current Law (KCL):
 KCL: Current that flows in is equal to current that flows out at each node.
- Note that there is a matrix C that connects currents and potential differences at the edges, and represent Ohm's law: y = Ce. We will talk about this later.

Graphs and networks: Kirchoff's Current Law (KCL) cont.

• The equation $A^T y = 0$ is Kirchoff's Current Law.

$$\begin{bmatrix} -1 & 0 & -1 & -1 & 0 \\ 1 & -1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & -1 \\ 0 & 0 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \end{bmatrix} = \mathbf{0}$$

1 4

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• The first equation refers to node one and indicates that the net current flow is zero. Similarly we get:

$$-y_{1} - y_{3} - y_{4} = 0$$

$$y_{1} - y_{2} = 0$$

$$y_{2} + y_{3} - y_{5} = 0$$

$$y_{4} + y_{5} = 0$$

Graphs and networks: Kirchoff's Current Law cont.

• The three solution vectors below, that satisfy Kirchoff's Current Law, represent total current running across the three possible loops.

$$\begin{bmatrix} 1 \\ 1 \\ -1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \\ -1 \\ 1 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 1 \\ 3 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 4 \\ 4 \\ 3 \end{bmatrix}$$

- We can see that the third solution (current running across the big external loop 3) is not independent from the first two solutions.
- The null space of *A^T* is, therefore, two dimensional, which is the same as the number of **small** loops. This is expectable also from the fact that:

$$\dim(N(A^T)) = m - r = 5 - 3 = 2$$

Graphs and networks: row space of incidence matrix

• Consider the columns space of A^T which is the row space of A.

$$A^{T} = \begin{bmatrix} -1 & 0 & -1 & -1 & 0 \\ 1 & -1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & -1 \\ 0 & 0 & 0 & 1 & 1 \end{bmatrix} \quad 2 \underbrace{1}_{2} \underbrace{1}_{3} \underbrace{4}_{5} \underbrace{4}_{3} \underbrace$$

- The pivot columns of *A*^{*T*} are located on the first, second and the fourth columns. These are associated with edges that form a graph without loops. As mentioned this graph is called a **tree**.
- The following relationships hold:
 - $\dim(N(A^T)) = m r \Rightarrow \#\text{small loops} = \#\text{edges} (\#\text{nodes} 1)$
 - #nodes #edges + #small loops = 1. This relationship is known as Euler's formula.

Real networks: Ohm's Law

- In real life networks we have: current=c ·potential differences.
- *c* is the so called **conductance**.
- It tells us how easily flow gets through an edge (high for metal, low for plastic etc.)
- Current through an edge is a function of the conductance across this edge **only** and the potential difference across the edge.
- Therefore, each edge is associated with a conductance.
- This yields y = Ce, with C being a diagonal matrix.
- The relationship y = Ce, is the so called **Ohm's Law**.

Summary

- In real life networks we have:
 - Potential differences: e = Ax
 - Ohm's Law: y = Ce
 - Kirchoff's Current Law: $A^T y = \mathbf{0}$
- The above three equations can be merged in a single equation as follows: $A^T y = \mathbf{0} \Rightarrow A^T C e = \mathbf{0} \Rightarrow A^T C A x = \mathbf{0}$