# EDISP (NWL2) <br> (English) Digital Signal Processing <br> Transform, FT, DFT 

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## Transform concept

We want to analyze the signal $\longrightarrow$ represent it as "built of" some buliding blocks (well known signals), possibly scaled

$$
x[n]=\sum_{k} A_{k} \phi_{k}[n]
$$

- The number $k$ of "blocks" $\phi_{k}[n]$ may be finite, infinite, or even a continuum (then $\Sigma \longrightarrow \int$ )
- Scaling coefficients $A_{k}$ are usually real or complex numbers
- $\phi_{k}$ are complex harmonics $e^{j \theta_{k} n}$ or cosines or wavelets ...
- If the representation (expansion) is unique for a class of functions, the set $\phi_{k}[n]$ is called a basis for this class.
- The above representation is an "Inverse ... transform". The ". . transform" (the forward one) is the way to calculate $A_{k}$ coefficients from the given signal $x[n]$.
- The forward transform is mathematically a cast onto the basis $\phi_{k}$, and it is calculated with inner product, scalar product of a signal with a dual basis $\tilde{\phi}_{k}$ functions $A_{k}=\left\langle x[n], \tilde{\phi}_{k}[n]\right\rangle$ (for an orthogonal transform, $\tilde{\phi}_{k}=\phi_{k}$ )
In a Fourier transform, we take the basis representing different frequencies.


## Fourier spectrum of a limited energy signal

$$
\sum_{n=-\infty}^{\infty}|x(n)|^{2}<\infty \quad,\left(x[n] \in \ell^{2}\right) \quad \begin{aligned}
& X\left(e^{j \theta}\right)-\text { a continuous, periodic func- } \\
& \text { tion. }
\end{aligned}
$$

Fourier spectrum definition:

$$
x(n)=\frac{1}{2 \pi} \int_{-\pi}^{\pi} x\left(e^{j \theta}\right) e^{j n \theta} d \theta
$$

$$
x\left(e^{j \theta}\right)=\sum_{n=-\infty}^{\infty} x(n) e^{-j n \theta}
$$

Linearity:

$$
a x[n]+b y[n] \stackrel{\mathcal{F}}{\longleftrightarrow} a X\left(e^{j \theta}\right)+b Y\left(e^{j \theta}\right)
$$

Time shift: $\quad x\left[n-n_{0}\right] \stackrel{\mathcal{F}}{\longleftrightarrow} e^{-j n_{0} \theta} X\left(e^{j \theta}\right)$,
Frequency shift: $e^{-j n \theta_{0}} x[n] \stackrel{\mathcal{F}}{\longleftrightarrow} X\left(e^{j\left(\theta-\theta_{0}\right)}\right)$
Convolution: $\quad x[n] * y[n] \stackrel{\mathcal{F}}{\longleftrightarrow} X\left(e^{j \theta}\right) \cdot Y\left(e^{j \theta}\right)$,
Modulation: $\quad x[n] \cdot y[n] \stackrel{\mathcal{F}}{\longleftrightarrow} \frac{1}{2 \pi} \int_{0}^{2 \pi} X\left(e^{j \phi}\right) \cdot Y\left(e^{j \theta-\phi}\right) d \phi$
(Parseval's): $\quad E=\sum_{n=-\infty}^{\infty}|x(n)|^{2}=\frac{1}{2 \pi} \int_{-\pi}^{\pi}\left|X\left(e^{j \theta}\right)\right|^{2} d \theta$

## Example

We sample $x_{a}(t)$ with $T_{s}=T / L$

$$
\begin{array}{cl}
x_{a}(t)=\left\{\begin{array}{ccc}
1 & \text { for } 0 \leq t<T \\
0 & \text { for } & \text { other } t
\end{array}\right. & x[n]=\left\{\begin{array}{cc}
1 & \text { for } n=0,1, \ldots, L-1 \\
0 & \text { for } \\
\text { other } n
\end{array}\right. \\
X_{a}(\omega)=\int_{-\infty}^{\infty} x_{a}(t) e^{-j \omega t} d t & x\left(e^{j \theta}\right)=\sum_{n=-\infty}^{\infty} x(n) e^{-j n \theta}
\end{array} \quad \begin{array}{ll}
X_{a}(\omega)=T \frac{\sin (\omega T / 2)}{\omega T / 2} e^{-j \omega T / 2} & x\left(e^{j \theta}\right)=e^{-j(L-1) \theta / 2} \frac{\sin (L \theta / 2)}{\sin (\theta / 2)} \\
& \text { (hint: } \left.\left(\sum_{n=0}^{N-1} q^{n}=\left(1-q^{N}\right) /(1-q)\right)\right)
\end{array}
$$

## Periodic (limited mean power) signal FT

The signal is periodic with period $\mathrm{N} \longrightarrow$ no component that is nonperiodic or periodic with different period.
Conclusion: only N -periodic components (this includes $\mathrm{N} / \mathrm{k}$ : $\mathrm{N} / 2, \mathrm{~N} / 3$, etc.) $\longrightarrow e^{j 2 \pi n k / N}$

$$
\frac{1}{N} \sum_{n=0}^{N-1}|x(n)|^{2}<\infty
$$

Fourier spectrum definition:

$$
x(n)=\frac{1}{N} \sum_{k=0}^{N-1} X(k) e^{j 2 \pi k n / N}
$$

$X(k)=\sum_{n=0}^{N-1} x(n) e^{-j 2 \pi k n / N}, \quad-\infty<k<\infty \quad \longrightarrow$ inverse transform
We represent $x[n]$ as a sum of $N$ complex discrete harmonics with angular frequencies $\theta_{k}=\frac{2 \pi}{N} \cdot k, \quad k=0,1, \ldots, N-1$

## 8 basis functions for $\mathrm{N}=8$ (real part only)

$\mathrm{k}=0 . .7$ basis functions for $\mathrm{N}=8$ (and the eight = zeroth)



## Example

$x_{p}[n]$ with period $N=10$ has $L=5$ nonzero samples
( $n=0,1, \ldots L-1$ )
$X(k)=\sum_{n=0}^{N-1} x_{p}(n) \mathrm{e}^{-\mathrm{j} 2 \pi k n / N}=\sum_{n=0}^{L-1} \mathrm{e}^{-\mathrm{j} 2 \pi k n / N}=\mathrm{e}^{-\mathrm{j}(L-1) \pi k / N} \frac{\sin (L \pi k / N)}{\sin (\pi k / N)}$,
The amplitude spectrum $|X[k]|=\left|\frac{\sin \left(L \theta_{k / 2}\right)}{\sin \left(\theta_{k} / 2\right)}\right|, \quad \theta_{k}=2 \pi k / N$ is shown

b)


## Discrete Fourier Transform

- A signal $x[n]$ defined for $-\infty<n<\infty$
- Its spectrum $X\left(e^{j \theta}\right)$ defined for continuous $0 \leq \theta<2 \pi$
- Life is short...
$\longrightarrow$ Let us take a fragment of $x[n]: x_{0}[n], n=0,1, \ldots, N-1$

$$
x_{0}[n]=x[n] g[n], \text { where } g[n]=\left\{\begin{array}{ccc}
1 & \text { for } & n=0,1, \ldots, N-1 \\
0 & \text { for } & \text { others } n
\end{array}\right.
$$

$g[n]$ - window (gate?) function (here: a rectangular window) ( $w[n]$ we reserve for white noise)
$\longrightarrow$ We take only $N$ values of $\theta_{k}=\frac{2 \pi}{N} k, \quad k=0,1, \ldots, N-1$

$$
x_{0}\left(\mathrm{e}^{\mathrm{j} \theta_{k}}\right)=\sum_{n=0}^{N-1} x_{0}(n) \mathrm{e}^{-\mathrm{j} n \theta_{k}}=\sum_{n=0}^{N-1} x_{0}(n) \mathrm{e}^{-\mathrm{j} 2 \pi n k / N}
$$

## DFT properties

Orthogonality - (see next slide)
Periodicity As we sample the spectrum, the reconstructed signal is periodic with period $N$. If we compute IDFT for $-\infty<n<\infty \ldots$

- A non-periodic signal was reconstructed as periodic
- A periodic signal was reconstructed as $N$-periodic

b)


## DFT as an orthogonal transform

An orthogonal transform (e.g. DFT) is a decomposition of a function (signal) on a set of orthogonal basis functions $\phi_{k}[n]$.

$$
x[n]=\frac{1}{N} \sum_{k=0}^{N-1} A(k) \cdot \phi_{k}[n]
$$

Because of $\phi_{k}[n]$ orthogonality, $A(k)$ are easy to calculate:

$$
A(k)=\sum_{n=0}^{N-1} x(n) \cdot \phi_{k}^{*}(n)
$$

Basis sequences (transform kernel) have to be orthogonal:
$\bar{N} \sum_{k=0}^{N-1} \phi_{k}(n) \cdot \phi_{m}^{*}(n)=\left\{\begin{array}{ll}1 & m=k \\ 0 & \text { otherwise }\end{array}\right.$ Scalar product is zero =orthogonal!
DFT basis functions $\phi_{k}(n)=\mathrm{e}^{-\mathrm{j} n \theta_{k}}=\mathrm{e}^{-\mathrm{j} 2 \pi n k / N}$ are orthogonal - we chose $\theta_{k}$ so it be!

## Inverse DFT

Let's take forward DFT definition as a linear equation set, with $x_{0}[n]$ as unknowns. When we multiply both sides by $\mathrm{e}^{\mathrm{j} 2 \pi r k / N}, r=0,1, \ldots, N-1$ and sum for $k=0,1, \ldots, N-1$

$$
\begin{gathered}
\sum_{k=0}^{N-1} x_{0}(k) \mathrm{e}^{\mathrm{j} 2 \pi r k / N}=\sum_{k=0}^{N-1}\left[\sum_{n=0}^{N-1} x_{0}(n) \mathrm{e}^{-\mathrm{j} 2 \pi n k / N}\right] \mathrm{e}^{\mathrm{j} 2 \pi r k / N}= \\
=\sum_{k=0}^{N-1} \sum_{n=0}^{N-1} x_{0}(n) \mathrm{e}^{\mathrm{j} 2 \pi k(r-n) / N}=\sum_{n=0}^{N-1} x_{0}(n) \sum_{k=0}^{N-1} \mathrm{e}^{\mathrm{j} 2 \pi k(r-n) / N} \\
\sum_{k=0}^{N-1} \mathrm{e}^{\mathrm{j} 2 \pi k(r-n) / N}=\left\{\begin{array}{rr}
N, & r=n \\
0, & r \neq n
\end{array} \Rightarrow \sum_{k=0}^{N-1} x_{0}(k) \mathrm{e}^{\mathrm{j} 2 \pi r k / N}=N x_{0}(r), \quad r=0,1\right. \\
x_{0}(n)=\frac{1}{N} \sum_{k=0}^{N-1} x_{0}(k) \mathrm{e}^{\mathrm{j} 2 \pi n k / N}, \quad n=0,1, \ldots, N-1
\end{gathered}
$$

## Forward and Inverse DFT - transformation matrix

$$
\begin{array}{rr}
\mathcal{F}(x)= & F \cdot x \\
F^{-1} \cdot \mathcal{F}(x)=F^{-1} \cdot F \cdot x \\
F^{-1} \cdot \mathcal{F}(x)= & x
\end{array}
$$

algebraic trivia:
How to construct $F$ matrix? $F_{k n}=e^{-j 2 \pi n k / N}$
What is $F^{-1}$ ? (not-so-trivial, but see IDFT slide)
Note nice properties of $F$ matrix...

