# FOURIER ANALYSIS \& METHODS 2020.02.21 

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#### Abstract

Caveat Emptor! These are just informal lecture notes. Errors are inevitable! Read at your own risk! Also, this is by no means a substitute for the textbook, which is warmly recommended: Fourier Analysis and Its Applications, by Gerald B. Folland. He was the first math teacher I had at university, and he is awesome. A brilliant writer. So, why am I even doing this? Good question...


## 1. The Sampling Theorem

This theorem is all about the interaction between Fourier series and Fourier coefficients and how to work with both simultaneously. It is one of the theory items, so its proof is important.

Theorem 1. Let $f \in L^{2}(\mathbb{R})$. We take the definition of the Fourier transform of $f$ to be

$$
\int_{\mathbb{R}} e^{-i x \xi} f(x) d x
$$

and we then assume that there is $L>0$ so that $\hat{f}(\xi)=0 \forall \xi \in \mathbb{R}$ with $|\xi|>L$. Then:

$$
f(t)=\sum_{n \in \mathbb{Z}} f\left(\frac{n \pi}{L}\right) \frac{\sin (n \pi-t L)}{n \pi-t L}
$$

## Proof:

Idea: Since the Fourier transform $\hat{f}$ has compact support, we can expand it as a Fourier series.
We therefore have

$$
\hat{f}(x)=\sum_{-\infty}^{\infty} c_{n} e^{i n \pi x / L}, \quad c_{n}=\frac{1}{2 L} \int_{-L}^{L} e^{-i n \pi x / L} \hat{f}(x) d x
$$

Idea: Use the FIT to express $f$ in terms of its Fourier transform.
We therefore have

$$
f(t)=\frac{1}{2 \pi} \int_{\mathbb{R}} e^{i x t} \hat{f}(x) d x=\frac{1}{2 \pi} \int_{-L}^{L} e^{i x t} \hat{f}(x) d x
$$

On the left we have used the fact that $\hat{f}$ is supported in the interval $[-L, L]$, thus the integrand is zero outside of this interval, so we can throw that part of the integral away.

Idea: Substitute the Fourier expansion of $\hat{f}$ into the integral.

So, we have

$$
f(t)=\frac{1}{2 \pi} \int_{-L}^{L} e^{i x t} \sum_{-\infty}^{\infty} c_{n} e^{i n \pi x / L} d x
$$

From here until the end of the proof, we will essentially just be computing. The coefficients

$$
c_{n}=\frac{1}{2 L} \int_{-L}^{L} e^{-i n \pi x / L} \hat{f}(x) d x=\frac{1}{2 L} \int_{\mathbb{R}} e^{i x(-n \pi / L)} \hat{f}(x) d x=\frac{2 \pi}{2 L} f\left(\frac{-n \pi}{L}\right)
$$

In the second equality we have used the fact that $\hat{f}(x)=0$ for $|x|>L$, so by including that part we don't change the integral. In the third equality we have used the FIT!!! So, we now substitute this into our formula above for

$$
f(t)=\frac{1}{2 \pi} \int_{-L}^{L} e^{i x t} \sum_{-\infty}^{\infty} \frac{\pi}{L} f\left(\frac{-n \pi}{L}\right) e^{i n \pi x / L} d x
$$

This is approaching the form we wish to have in the theorem, but the argument of the function $f$ has a pesky negative sign. That can be remedied by switching the order of summation, which does not change the sum, so

$$
f(t)=\frac{1}{2 L} \int_{-L}^{L} e^{i x t} \sum_{-\infty}^{\infty} f\left(\frac{n \pi}{L}\right) e^{-i n \pi x / L} d x
$$

We may also interchange the summation with the integra $\sqrt[1]{1}$

$$
f(t)=\frac{1}{2 L} \sum_{-\infty}^{\infty} f\left(\frac{n \pi}{L}\right) \int_{-L}^{L} e^{x(i t-i n \pi / L)} d x
$$

We then compute

$$
\int_{-L}^{L} e^{x(i t-i n \pi / L)} d x=\frac{e^{L(i t-i n \pi / L)}}{i(t-n \pi / L)}-\frac{e^{-L(i t-i n \pi / L)}}{i(t-n \pi / L)}=\frac{2 i}{i(t-n \pi / L)} \sin (L t-n \pi)
$$

Substituting,

$$
f(t)=\sum_{-\infty}^{\infty} f\left(\frac{n \pi}{L}\right) \frac{\sin (L t-n \pi)}{L t-n \pi}
$$

Of course my dyslexia has ended up with things being backwards, but it is not a problem because sine is odd so

$$
\sin (L t-n \pi)=-\sin (n \pi-L t)
$$

SO

$$
\frac{\sin (L t-n \pi)}{L t-n \pi}=\frac{-\sin (n \pi-L t)}{L t-n \pi}=\frac{\sin (n \pi-L t}{n \pi-L t}
$$

[^0]Today we shall investigate some transforms related to the Fourier transform. The first two can be used to solve PDEs on half lines, if the boundary condition is suitable.
2.1. Motivation: heat equation on a semi-infinite rod with an insulated end. We have found ourselves in possession of a giant rod which is insulated at the one end and goes out to infinity at the other. It has an initial temperature distribution given by a function $f(x)$ which is bounded, continuous and an element of $\mathcal{L}^{2}$. We therefore wish to solve the problem:

$$
u_{t}-u_{x x}=0, \quad u_{x}(0, t)=0, \quad u(x, 0)=f(x), \quad x \in[0, \infty)
$$

To solve such a problem we will use a Fourier cosine transform together with the Fourier cosine transform inverse theorem.

### 2.2. Fourier sine and cosine transforms and their inverse formulas.

Definition 2. Let $f$ be in $\mathcal{L}^{1}$ or $\mathcal{L}^{2}$ on $(0, \infty)$. The Fourier cosine transform,

$$
\mathcal{F}_{c}(f)(\xi):=\int_{0}^{\infty} f(x) \cos (\xi x) d x
$$

The Fourier sine transform,

$$
\mathcal{F}_{s}(f)(\xi):=\int_{0}^{\infty} f(x) \sin (\xi x) d x
$$

As with the Fourier transform, the Fourier sine and cosine transforms also have inversion formula.

Theorem 3. Assume that $f \in \mathcal{L}^{2}[0, \infty)$. Then we have the Fourier cosine inversion formula

$$
f(x)=\frac{2}{\pi} \int_{0}^{\infty} \mathcal{F}_{c}(f)(\xi) \cos (x \xi) d \xi
$$

We also have the Fourier sine inversion formula

$$
f(x)=\frac{2}{\pi} \int_{0}^{\infty} \sin (x \xi) \mathcal{F}_{s}(f)(\xi) d \xi
$$

Proof: First, let us extend $f$ evenly to $\mathbb{R}$, denoting this extension by $f_{e}$, so that $f_{e}(-x)=f_{e}(x)$. We compute the standard Fourier transform:
$\hat{f}_{e}(\xi)=\int_{\mathbb{R}} f_{e}(x) e^{-i x \xi} d x=\int_{\mathbb{R}} f_{e}(x)(\cos (x \xi)-i \sin (x \xi)) d x=2 \int_{0}^{\infty} f(x) \cos (x \xi) d x$.
The term with the sine has dropped out because $f_{e}(x) \sin (x \xi)$ is an odd function of $x$. The term with the cosine gets doubled because $f_{e}(x) \cos (x \xi)$ is an even function. So, all together we have computed:

$$
\hat{f}_{e}(\xi)=2 \int_{0}^{\infty} f(x) \cos (x \xi) d x=2 \mathcal{F}_{c}(f)(\xi)
$$

Since the cosine is an even function,

$$
\hat{f}_{e}(\xi)=\hat{f}_{e}(-\xi)
$$

So, we also have that $\mathcal{F}_{c}(f)$ is an even function. The inversion formula (FIT) says that

$$
\begin{gathered}
f_{e}(x)=\frac{1}{2 \pi} \int_{\mathbb{R}} e^{i x \xi} \hat{f}_{e}(\xi) d \xi=\frac{1}{\pi} \int_{\mathbb{R}} e^{i x \xi} \mathcal{F}_{c}(f)(\xi) d \xi \\
=\frac{1}{\pi} \int_{\mathbb{R}}(\cos (x \xi)+i \sin (x \xi)) \mathcal{F}_{c}(f)(\xi) d \xi=\frac{2}{\pi} \int_{0}^{\infty} e^{i x \xi} \mathcal{F}_{c}(f)(\xi) d \xi
\end{gathered}
$$

This is the cosine-FIT! Above we have used the fact that $\mathcal{F}_{c}(f)$ is an even function. Hence its product with the cosine is also an even function, so that part of the integral gets a factor of two when we integrate only over the positive real line. The product of an even function like $\mathcal{F}_{c}(f)$ with an odd function, like the sine, is odd, so that integral vanishes.

On the other hand, we may also define the odd extension, $f_{o}$ which satisfies $f_{o}(-x)=-f_{o}(x)($ for $x \neq 0)$. The value of $f$ at zero is not really important at this moment $\left[^{2}\right.$ We compute the standard Fourier transform of the odd extension:

$$
\begin{gathered}
\hat{f}_{o}(\xi)=\int_{\mathbb{R}} f_{o}(x) e^{-i x \xi} d x=\int_{\mathbb{R}} f_{o}(x)(\cos (x \xi)-i \sin (x \xi)) d x=-2 i \int_{0}^{\infty} f(x) \sin (x \xi) d x \\
=-2 i \mathcal{F}_{s}(f)(\xi)
\end{gathered}
$$

Above, we have used the fact that $f_{o}$ is odd, and therefore so is its product with the cosine. On the other hand, the product with the sine is an even function, which explains the factor of 2 . Since the sine itself is odd, we have that $\hat{f}_{o}$ is an odd function and similarly $\mathcal{F}_{s}(f)(\xi)$ is also an odd function. We apply the FIT:

$$
\begin{aligned}
f_{o}(x) & =\frac{1}{2 \pi} \int_{\mathbb{R}} e^{i x \xi} \hat{f}_{o}(\xi) d \xi=-\frac{i}{\pi} \int_{\mathbb{R}}(\cos (x \xi)+i \sin (x \xi)) \mathcal{F}_{s}(f)(\xi) d \xi \\
& \left.=\frac{1}{\pi} \int_{\mathbb{R}} \sin (x \xi) \mathcal{F}_{s}(f)(\xi) d \xi=\frac{2}{\pi} \int_{0}^{\infty} \sin (x \xi) \mathcal{F}_{s}(f)(\xi) d \xi\right)
\end{aligned}
$$

This is the sine-FIT! Above we have used the fact that $\mathcal{F}_{s}(f)$ is an odd function, and therefore so is its product with the cosine. On the other hand the product of two odd functions is an even function, so that is the reason for the factor of 2 .

### 2.3. Solving the heat equation on a semi-infinite rod with insulated end.

 We wish to solve the problem:$$
u_{t}-u_{x x}=0, \quad u_{x}(0, t)=0, \quad u(x, 0)=f(x), \quad x \in[0, \infty)
$$

Assume that by some method, we have obtained a solution $u(x, t)$ defined on $[0, \infty)_{x} \times[0, \infty)_{t}$. To see if we may use a Fourier sine or cosine transform method, let us see what happens when we extend our solution evenly or oddly. The even extension would satisfy, by the cosine-FIT:

$$
u_{e}(x, t)=\frac{2}{\pi} \int_{0}^{\infty} \mathcal{F}_{c}(u)(\xi) \cos (x \xi) d \xi
$$

The odd extension would satisfy, by the sine-FIT

$$
u_{o}(x, t)=\frac{2}{\pi} \int_{0}^{\infty} \sin (x \xi) \mathcal{F}_{s}(f)(\xi) d \xi
$$

[^1]OBS! The extension matches up with our original function on the positive real line (that is how an extension works!) We need the derivative with respect to $x$ to vanish at $x=0$. Let's just differentiate these expressions. Note that the $x$ dependence is only in the sine or cosine term so we have:

$$
\partial_{x} u_{e}(x, t)=-\frac{2}{\pi} \int_{0}^{\infty} \mathcal{F}_{c}(u)(\xi) \xi \sin (x \xi) d \xi \Longrightarrow \partial_{x} u_{e}(0, t)=0
$$

On the other hand
$\partial_{x} u_{o}(x, t)=\frac{2}{\pi} \int_{0}^{\infty} \xi \cos (x \xi) \mathcal{F}_{s}(u)(\xi) d \xi \Longrightarrow \partial_{x} u_{o}(0, t)=\frac{2}{\pi} \int_{0}^{\infty} \xi \mathcal{F}_{s}(u)(\xi) d \xi=? ? ?$
The even extension automatically gives us the desired boundary condition whereas the odd extension leads to something complicated looking, which we have no reason to know is zero.

This indicates that we have a decent chance of solving the problem by:
(1) Extending the initial data evenly to the real line.
(2) Solving the problem using the Fourier transform on the real line.
(3) Verifying that the solution satisfies all the conditions: the PDE, the IC, and the BC.
We do this. Extend $f$ evenly, and write the extension as $f_{e}$. Then the solution to the homogeneous heat equation on the real line with initial data $f_{e}$ is

$$
u_{e}(x, t)=\frac{1}{2 \sqrt{\pi t}} \int_{\mathbb{R}} f_{e}(y) e^{-\frac{(x-y)^{2}}{4 t}} d y
$$

We split up the integral:

$$
\begin{aligned}
& \int_{-\infty}^{0} f_{e}(y) e^{-(x-y)^{2} /(4 t)} d y+\int_{0}^{\infty} f_{e}(y) e^{-(x-y)^{2} /(4 t)} d y \\
= & -\int_{\infty}^{0} f_{e}(z) e^{-(z+x)^{2} /(4 t)} d z+\int_{0}^{\infty} f_{e}(y) e^{-(x-y)^{2} /(4 t)} d y
\end{aligned}
$$

Above we made the substitution that $z=-y$ in the first integral. Due to the evenness of $f_{e}$, nothing happens when we change $y=-z$. Reversing the limits of integration we obtain

$$
-\int_{\infty}^{0} f_{e}(z) e^{-(z+x)^{2} /(4 t)} d z=\int_{0}^{\infty} f_{e}(z) e^{-(z+x)^{2} /(4 t)} d z=\int_{0}^{\infty} f_{e}(y) e^{-(x+y)^{2} /(4 t)} d y
$$

So, all together we have

$$
u_{e}(x, t)=\frac{1}{2 \sqrt{\pi t}} \int_{0}^{\infty} f(y)\left(e^{-\frac{(x-y)^{2}}{4 t}}+e^{-\frac{(x+y)^{2}}{4 t}}\right) d y
$$

Is this an even function? Let us verify:

$$
e^{-\frac{(x-y)^{2}}{4 t}}+e^{-\frac{(x+y)^{2}}{4 t}}=e^{-\frac{(-x-y)^{2}}{4 t}}+e^{-\frac{(-x+y)^{2}}{4 t}} .
$$

AWESOME! Our solution to the heat equation in this way is EVEN. Therefore, it is the same on the left and right sides. So, we can simply let

$$
u(x, t)=u_{e}(x, t)=\frac{1}{2 \sqrt{\pi t}} \int_{0}^{\infty} f(y)\left(e^{-\frac{(x-y)^{2}}{4 t}}+e^{-\frac{(x+y)^{2}}{4 t}}\right) d y
$$

The way we have built it, it satisfies the IC, BC, and the PDE!

Exercise 1. Solve:

$$
u_{t}-u_{x x}=0, \quad u(0, t)=0, \quad u(x, 0)=\phi(x), \quad x \in[0, \infty)
$$

Above, we assume that $\phi$ is bounded, continuous, and in $\mathcal{L}^{2}$. Hint: extend $\phi$ oddly this time, and use the Fourier sine inverse theorem.
2.4. Discrete and fast Fourier transform. We have seen that computing the Fourier transform is not the easiest thing in the world. The example with the Gaussian involving all those tricks: completing the square, complex analysis and contour integral is a nice and easy case. However, in the real world you may come across functions and not know how to compute the Fourier transform by hand, nor be able to find it in BETA. It could be lurking in one of our giant handbooks of calculations (Abramowitz \& Stegun, Gradshteyn \& Rhizik, to name a few). Or it could simply never have been computed analytically. In this case you may compute something called the discrete Fourier transform.

We start with a function, $f(t)$, and think of analyzing $f(t)$ as time analysis, whereas analyzing $\hat{f}(\xi)$ as frequency analysis. We shall consider a finite dimensional Hilbert space:

$$
\mathbb{C}^{N}=\left\{\left(s_{n}\right)_{n=0}^{N-1}, \quad s_{n} \in \mathbb{C}, \quad\left\langle\left(s_{n}\right),\left(t_{n}\right)\right\rangle:=\sum_{n=0}^{N-1} s_{n} \overline{t_{n}}\right\}
$$

Now let

$$
e_{k}(n):=\frac{e^{2 \pi i k n / N}}{\sqrt{N}}
$$

Proposition 4. Let

$$
e_{k}:=\left(e_{k}(n)\right)_{n=0}^{N-1}
$$

Then

$$
\left\{e_{k}\right\}_{k=0}^{N-1}
$$

are an ONB of $\mathbb{C}^{N}$.
Proof: We simply compute. It is so cute and discrete!

$$
\left\langle e_{k}, e_{j}\right\rangle=\frac{1}{N} \sum_{n=0}^{N-1} e^{2 \pi i k n / N} e^{-2 \pi i j n / N}=\frac{1}{N} \sum_{n=0}^{N-1} e^{2 \pi i(k-j) n / N}
$$

If $j=k$ the terms are all 1 , and so the total is $N$ which divided by $N$ gives 1 . Otherwise, we may without loss of generality assume that $k>j$ (swap names if not the case). Then we are staring at a geometric series! We know how to sum it

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

Here it is super important that $k-j$ is a number between 1 and $N-1$. We know this because $0 \leq j<k \leq N-1$. Hence when we subtract $j$ from $k$, we get something between 1 and $N-1$. So we are not dividing by zero!

Now we shall fix $T$ small and $N$ large and look at $f(t)$ just on the interval $[0,(N-1) T]$. Let

$$
f\left(t_{n}\right):=f(n T), \quad t_{n}=n T
$$

Basically, we're going to identify $f$ with an element of $\mathbb{C}^{N}$, namely

$$
\left(f\left(t_{n}\right)\right)_{n=0}^{N-1} .
$$

Definition 5. Let

$$
w_{k}:=\frac{2 \pi k}{N T}
$$

The discrete Fourier transform of $f$ at $w_{k}$ is defined to be

$$
F\left(w_{k}\right):=\left\langle\left(f\left(t_{n}\right)\right), e_{k}\right\rangle=\sum_{n=0}^{N-1} \frac{f\left(t_{n}\right) e^{-2 \pi i k n / N}}{\sqrt{N}}
$$

This can also be written as

$$
\sum_{n=0}^{N-1} \frac{f\left(t_{n}\right) e^{-i w_{k} t_{n}}}{\sqrt{N}}
$$

Example 1. One of the fun facts about the discrete Fourier transform is that we can Fourier transform functions which are neither in $\mathcal{L}^{2}$ nor in $\mathcal{L}^{1}$. For example, let's compute the discrete Fourier transform of

$$
f(x)=x, \quad T=\frac{1}{10}, \quad N=5
$$

So, we identify $f$ with the vector

$$
(0,0.1,0.2,0.3,0.4)
$$

Then,

$$
F\left(w_{k}\right):=\sum_{n=0}^{4} \frac{n e^{-2 \pi i k n / 5}}{10 \sqrt{5}}
$$

So, we identify the Fourier transform of $f$ with the vector

$$
\left(\sum_{n=0}^{4} \frac{n}{10 \sqrt{5}}, \sum_{n=0}^{4} \frac{n e^{-2 \pi i n / 5}}{10 \sqrt{5}}, \sum_{n=0}^{4} \frac{n e^{-4 \pi i n / 5}}{10 \sqrt{5}}, \sum_{n=0}^{4} \frac{n e^{-6 \pi i n / 5}}{10 \sqrt{5}}, \sum_{n=0}^{4} \frac{n e^{-8 \pi i n / 5}}{10 \sqrt{5}}\right) .
$$

Proposition 6. We have the inversion formula

$$
f\left(t_{n}\right)=\sum_{k=0}^{N-1} F\left(w_{k}\right) e_{n}(k)=\left\langle\left(F\left(w_{k}\right)\right), \bar{e}_{n}\right\rangle
$$

Proof: We simply compute. By definition

$$
\left\langle\left(F\left(w_{k}\right)\right), \bar{e}_{n}\right\rangle=\sum_{k=0}^{N-1} F\left(w_{k}\right) e_{n}(k)
$$

Now, we insert the definition of $F\left(w_{k}\right)$ which gives us another sum, so we use a different index there. Hence we have

$$
\begin{aligned}
& \sum_{k=0}^{N-1} \sum_{m=0}^{N-1} \frac{f\left(t_{m}\right) e^{-i w_{k} t_{m}}}{\sqrt{N}} \frac{e^{2 \pi i k n / N}}{\sqrt{N}}=\frac{1}{N} \sum \sum f\left(t_{m}\right) e^{-2 \pi i k m / N} e^{2 \pi i k n / N} \\
& \quad=\frac{1}{N} \sum \sum f\left(t_{m}\right) e^{2 \pi i k(n-m) / N}=\frac{1}{N} \sum_{m=0}^{N-1} f\left(t_{m}\right) \sum_{k=0}^{N-1} e^{2 \pi i k(n-m) / N}
\end{aligned}
$$

$$
=\sum_{m=0}^{N-1} f\left(t_{m}\right) \sum_{k=0}^{N-1} \frac{e^{-2 \pi i k m / N}}{\sqrt{N}} \frac{\overline{e^{-2 \pi i k n / N}}}{\sqrt{N}}=\sum_{m=0}^{N-1} f\left(t_{m}\right)\left\langle e_{m}, e_{n}\right\rangle .
$$

By the proposition we just proved before,

$$
\left\langle e_{m}, e_{n}\right\rangle=\delta_{n, m}= \begin{cases}0 & n \neq m \\ 1 & n=m\end{cases}
$$

So, the only term which survives is when $m=n$, and so we get

$$
f\left(t_{n}\right)
$$

Example 2. Now, let's see if the inversion formula actually works for our example... First, we should have

$$
\begin{aligned}
\sum_{k=0}^{4} F\left(w_{k}\right) e_{0}(k) & =\sum_{k=0}^{4} \sum_{n=0}^{4} \frac{n e^{-2 \pi i k n / 5}}{10 \sqrt{5}} \frac{1}{\sqrt{5}} \\
=\frac{1}{50} \sum_{n=0}^{4} n \sum_{k=0}^{4} e^{-2 \pi i k n / 5} & =\frac{1}{50} \sum_{n=1}^{4} \frac{1-e^{-2 \pi i n}}{1-e^{-2 \pi i n / 5}}=0=f\left(t_{0}\right) .
\end{aligned}
$$

Let's try another value:

$$
\begin{gathered}
\sum_{k=0}^{4} F\left(w_{k}\right) e_{1}(k)=\sum_{k=0}^{4} \sum_{m=0}^{4} \frac{m e^{-2 \pi i k m / 5}}{10 \sqrt{5}} \frac{e^{2 \pi i k / 5}}{\sqrt{5}} \\
=\frac{1}{50} \sum_{n=1}^{4} n \sum_{k=0}^{4} e^{-2 \pi i k(n-1) / 5}
\end{gathered}
$$

For $n=2,3,4$, the sum over $k$ gives

$$
\frac{1-e^{-2 \pi i(n-1)}}{1-e^{-2 \pi i(n-1) / 5}}=0
$$

For $n=1$, the sum over $k$ gives 5 . Thus, the only term that survives is the term with $n=1$, for which we obtain

$$
\frac{1}{50}(1)(5)=\frac{1}{10}=f\left(t_{1}\right)
$$

So, it is indeed working as it should. This is rather tedious, however.
Now, we can see this as matrix multiplication. In the discrete Fourier transform, we sampled $f$ at the finitely many points $t_{0}, \ldots, t_{N-1}$. We therefore identify $f$ with a vector

$$
\left[\begin{array}{c}
f\left(t_{0}\right) \\
f\left(t_{1}\right) \\
\vdots \\
f\left(t_{N-1}\right)
\end{array}\right] .
$$

Similarly, the Fourier transform can be identified with the vector:

$$
\left[\begin{array}{c}
F\left(w_{0}\right) \\
F\left(w_{1}\right) \\
\cdots \\
F\left(w_{N-1}\right)
\end{array}\right]
$$

This vector is the product of the matrix

$$
\left[\begin{array}{lll}
\bar{e}_{0} & \bar{e}_{1} & \ldots \bar{e}_{N-1}
\end{array}\right]
$$

whose columns are

$$
\bar{e}_{n}=\frac{1}{\sqrt{N}}\left[\begin{array}{c}
e^{0} \\
e^{-2 \pi i n / N} \\
e^{-2 \pi i(2) n / N} \\
\ldots e^{-2 \pi i k n / N} \\
\ldots \\
e^{-2 \pi i n(N-1) / N}
\end{array}\right]
$$

together with the vector

$$
\left[\begin{array}{c}
f\left(t_{0}\right) \\
f\left(t_{1}\right) \\
\cdots \\
f\left(t_{N-1}\right)
\end{array}\right]
$$

That is

$$
\left[\begin{array}{c}
F\left(w_{0}\right) \\
F\left(w_{1}\right) \\
\ldots \\
F\left(w_{N-1}\right)
\end{array}\right]=\left[\begin{array}{lll}
\bar{e}_{0} & \bar{e}_{1} & \ldots \bar{e}_{N-1}
\end{array}\right]\left[\begin{array}{c}
f\left(t_{0}\right) \\
f\left(t_{1}\right) \\
\ldots \\
f\left(t_{N-1}\right)
\end{array}\right]
$$

This entails a LOT of calculations. We can speed it up by being clever. Many calculations are repeated in fact. Assume that $N=2^{X}$ for some giant power $X$. The idea is to split up into even and odd terms. We do this:

$$
F\left(w_{k}\right)=\frac{1}{\sqrt{N}}\left[\sum_{j=0}^{\frac{N}{2}-1} f\left(t_{2 j}\right) e^{-2 \pi i k(2 j) / N}+\sum_{j=0}^{\frac{N}{2}-1} f\left(t_{2 j+1}\right) e^{-2 \pi i k(2 j+1) / N}\right]
$$

We introduce the slightly cumbersome notation:

$$
e_{N}^{k}(n)=e^{-2 \pi i k n / N}
$$

Then,

$$
e_{N}^{k}(2 j)=e^{-2 \pi i k(2 j) / N}=e^{-2 \pi i k j /(N / 2)}=e_{N / 2}^{k}(j) .
$$

Now we only need an $\frac{N}{2} \times \frac{N}{2}$ matrix! You see, writing this way,

$$
F\left(w_{k}\right)=\frac{1}{\sqrt{N}}\left[\sum_{j=0}^{\frac{N}{2}-1} f\left(t_{2 j}\right) e_{N / 2}^{k}(j)+e_{N}^{k}(1) \sum_{j=0}^{\frac{N}{2}-1} f\left(t_{2 j+1}\right) e_{N / 2}^{k}(j)\right]
$$

We can repeat this many times because $N$ is a power of 2 . We just keep chopping in half. If we do this as many times as possible, we will need to do on the order of

$$
\frac{N}{2} \log _{2}(N)
$$

computations. This is in comparison to the original method which had an $N \times N$ matrix and was thus on the order of $N^{2}$ computations. For example, if $N=2^{10}$, then comparing $N^{2}=2^{20}$ to $\frac{N}{2} \log _{2} N=2^{9} * 10$, we see that

$$
\frac{2^{10} * 5}{2^{20}}=\frac{x}{100} \Longrightarrow 100 * 2^{10} * 5=2^{20} x \Longrightarrow 2^{2} * 5^{3} * 2^{10} 2^{-20}=x
$$

$$
5^{3} 2^{-8}=x \approx 0.488
$$

This means the amount of work we are doing by using the FFT is less than $0.5 \%$ of the work done using the standard DFT. In other words, we save over $99.5 \%$ by doing the FFT. That's why it's called FAST.

### 2.5. Answers to the exercises to be done oneself.

(1) (Eö 9) Compute (with help of Fourier transform)

$$
\int_{\mathbb{R}} \frac{\sin (x)}{x\left(x^{2}+1\right)} d x
$$

(This is in the back of the EÖ document!)
(2) (EÖ 67) Compute the Fourier transform of the characteristic function for the interval $(a, b)$ both directly and by using the known case for the interval $(-a, a)$. (This is in the back of the EÖ document!)
(3) (7.2.8) Given $a>0$ let $f(x)=e^{-x} x^{a-1}$ for $x>0, f(x)=0$ for $x \leq 0$. Show that $\hat{f}(\xi)=\Gamma(a)(1+i \xi)^{-a}$ where $\Gamma$ is the Gamma function.

Well, there are not really answers to make sense of here. My hint was to do a substitution of variables:

$$
\hat{f}(\xi)=\int_{0}^{\infty} e^{-i x \xi-x} x^{a-1} d x
$$

On the other hand

$$
\Gamma(a)=\int_{0}^{\infty} e^{-t} t^{a-1} d t
$$

So let's try making

$$
x(1+i \xi)=t \Longrightarrow d x(1+i \xi)=d t \Longrightarrow \frac{d t}{1+i \xi}=d x
$$

Our integral becomes

$$
\begin{aligned}
\hat{f}(\xi) & =\int_{0}^{(1+i \xi) \infty} e^{-t}\left(\frac{t}{1+i \xi}\right)^{a-1} \frac{d t}{1+i \xi} \\
& =(1+i \xi)^{-a} \int_{0}^{(1+i \xi) \infty} e^{-t} t^{a-1} d t
\end{aligned}
$$

We integrate along the line from 0 to $(1+i \xi) R=R+i R \xi$. For $\xi>0$ that is the first diagonal bit. Next, integrate from $R+i R \xi$ to $R$. The integrate back along the real axis from $R$ to zero. Our integrand is $e^{-z} z^{a-1}$. Inside the triangle it's holomorphic. So by complex analysis the integral around the triangle is zero. Since $\left|e^{-z}\right|=e^{-x}$ if $z=x+i y$ for $x, y \in \mathbb{R}$, along the right side of the triangle the integral is super small, tending to zero. That says the the integral along this funny diagonal line and the integral going from $R$ to 0 are tending to be equal. More precisely $\lim _{R \rightarrow \infty} \int_{0}^{R(1+i \xi)} f(z) d z+\int_{R}^{0} f(z) d z=0$. Hence since flipping the integral changes its sign $\lim _{R \rightarrow \infty} \int_{0}^{R(1+i \xi)} f(z) d z=\int_{0}^{\infty} f(z) d z$. So
$\hat{f}(\xi)=(1+i \xi)^{-a} \int_{0}^{(1+i \xi) \infty} e^{-t} t^{a-1} d t=(1+i \xi)^{-a} \int_{0}^{\infty} e^{-t} t^{a-1} d t$.
This is $(1+i \xi)^{-a} \Gamma(a)$.
(4) (7.2.12) For $a>0$ let

$$
f_{a}(x)=\frac{a}{\pi\left(x^{2}+a^{2}\right)}, \quad g_{a}(x)=\frac{\sin (a x)}{\pi x}
$$

Use the Fourier transform to show that: $f_{a} * f_{b}=f_{a+b}$ and $g_{a} * g_{b}=g_{\min (a, b)}$.
So we transform:

$$
\widehat{f_{a} * f_{b}}(\xi)=\widehat{f}_{a}(\xi) \widehat{f}_{b}(\xi)=e^{-a|\xi|-b|\xi|}=e^{-(a+b) \mid \xi}
$$

Now we use the FIT to say:

$$
f_{a} * f_{b}(x)=\frac{1}{2 \pi} \int_{\mathbb{R}} e^{-(a+b)|\xi|} e^{i x \xi} d \xi
$$

OBS! The integral on the right side this is the Fourier transform of $e^{-(a+b)|\xi|}$ at the point $-x$ rather than $x$. So we use our beloved Table 2 (item 11) to say that the Fourier transform of this function at the point $-x$ is

$$
2(a+b)\left(x^{2}+(a+b)^{2}\right)^{-1}
$$

so substituting

$$
f_{a} * f_{b}(x)=\frac{1}{2 \pi} 2(a+b)\left(x^{2}+(a+b)^{2}\right)^{-1}=\frac{(a+b)}{\pi\left(x^{2}+(a+b)^{2}\right)}=f_{a+b}(x)
$$

We do the same trick to solve the $g$ exercise, yo.

$$
\widehat{g_{a} * g_{b}}(\xi)=\widehat{g_{a}}(\xi) \widehat{g_{b}}(\xi)=\chi_{a}(\xi) \chi_{b}(\xi)=\chi_{\min (a, b)}(\xi)
$$

The last step follows from the the definition of the characteristic function. So, we use the FIT again to say:

$$
g_{a} * g_{b}(x)=\frac{1}{2 \pi} \int_{\mathbb{R}} e^{i x \xi} \chi_{\min (a, b)}(\xi) d \xi
$$

Same trick: integral on the right is the Fourier transform of $\chi_{\min (a, b)}$ at the point $-x$ (rather than $x$ ). So we use our favorite Table 2 to say that
$g_{a} * g_{b}(x)=\frac{1}{2 \pi} x^{-1} 2 \sin (\min (a, b) x)=\frac{\sin (\min (a, b) x)}{\pi x}=g_{\min (a, b)}(x)$.
(5) (Eö 6.d,e) Compute the Fourier transform of:

$$
e^{-a|t|} \sin (b t), \quad(a, b>0), \quad \frac{t}{t^{2}+2 t+5}
$$

(This is in the back of the EÖ document!)
(6) (EÖ 15) Find a solution to the equation

$$
u(t)+\int_{-\infty}^{t} e^{\tau-t} u(\tau) d \tau=e^{-2|t|}
$$

(This is in the back of the EÖ document!)
(7) (Eö 11) For the function

$$
f(t)=\int_{0}^{2} \frac{\sqrt{w}}{1+w} e^{i w t} d w
$$

compute

$$
\int_{\mathbb{R}} f(t) \cos (t) d t, \quad \int_{\mathbb{R}}|f(t)|^{2} d t
$$

(This is in the back of the EÖ document!)

## References

[1] Gerald B. Folland, Fourier Analysis and Its Applications, Pure and Applied Undergraduate Texts Volume 4, (1992).


[^0]:    ${ }^{1}$ None of this makes sense pointwise; we are working over $L^{2}$. The key property which allows interchange of limits, integrals, sums, derivatives, etc is absolute convergence. This is the case here because elements of $L^{2}$ have $\int|f|^{2}<\infty$. That is precisely the type of absolute convergence required.

[^1]:    ${ }^{2}$ This is because we are working in $\mathcal{L}^{2}$ which ignores sets of measure zero, and a single point is a set of measure zero.

