



***If you can't choose, throw dice
– Random constructions in
mathematical (harmonic) analysis***

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Harmonic analysis

(traditionally) = Fourier analysis

**= study of decomposition ('analysis') of
functions / signals into basic sine waves
(‘harmonics’)**

$$f(x) = \sum_{k=-\infty}^{\infty} \hat{f}(k) e^{i2\pi kx}$$



Harmonic analysis

(today) = study of multiple phenomena that have emerged and grown ‘around’ the classical theory of Fourier series

- * interactions of spatio-temporal and spectral information**
- * cancellations, orthogonality, oscillations**



'Modern' block waves and wavelets: Rademacher and Haar functions

$$r_j(x) := \text{sign}(\sin(2\pi \cdot 2^j x))$$

$$h_{j,k}(x) := h_I(x) := \frac{1_I(x)}{|I|^{1/2}} r_j(x)$$

$$I = I_{j,k} = \left[\frac{k}{2^j}, \frac{k+1}{2^j} \right)$$



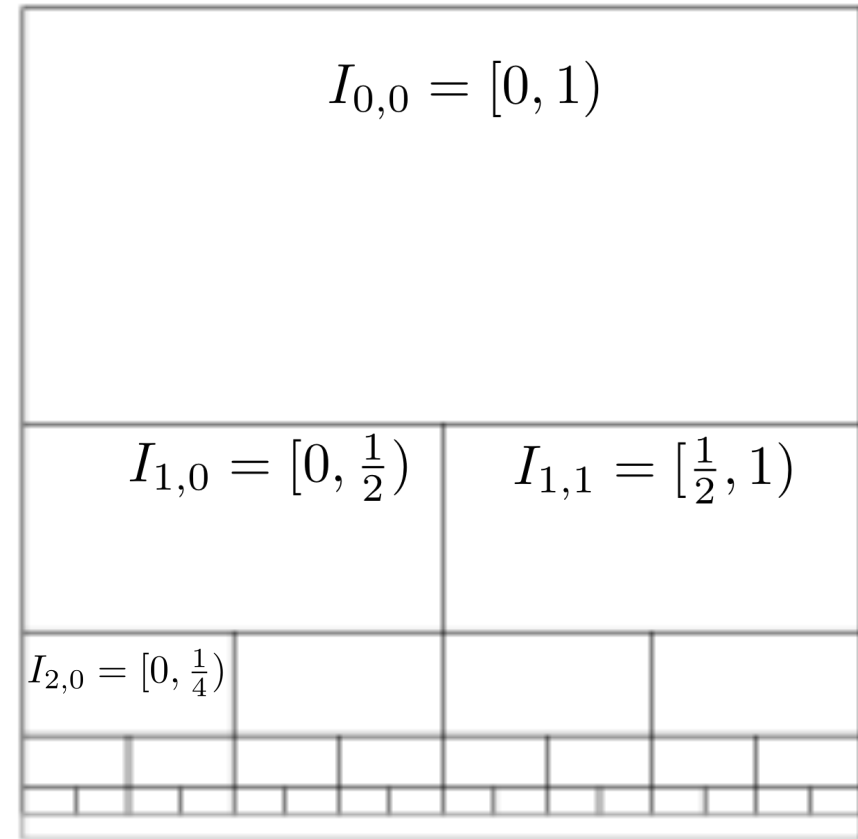
Dyadic analysis of a function $f(x)$

- Refine grid repeatedly dividing into halves
- At every scale, replace f by its average on each dyadic interval
- Change from one scale to next:

$j = 0$

$j = 1$

$j = 2$



$$\Delta_I f := 1_{I_{\text{left}}} \langle f \rangle_{I_{\text{left}}} + 1_{I_{\text{right}}} \langle f \rangle_{I_{\text{right}}} - 1_I \langle f \rangle_I = h_I \langle h_I, f \rangle$$

$$f = \sum_I \Delta_I f = \sum_I h_I \langle h_I \rangle f$$



More robust for 'rough' situations

$$\Delta_I^\mu f := 1_{I_{\text{left}}} \langle f \rangle_{I_{\text{left}}}^\mu + 1_{I_{\text{right}}} \langle f \rangle_{I_{\text{right}}}^\mu - 1_I \langle f \rangle_I^\mu = h_I^\mu \langle h_I^\mu, f \rangle$$

- **Averaging over intervals is meaningful with respect to any mass distribution ('measure') in place of the uniform distribution (Lebesgue measure)**

$$\langle f \rangle_I^\mu := \frac{1}{\mu(I)} \int_I f(x) d\mu(x)$$

- **Weighted Haar functions h_I^μ remain piecewise constant, average to zero with respect to μ**



Dyadic analysis – what for?

- **Divide and conquer – estimates for general f reduced to simpler $\Delta_I f$.**

$$f = \sum_I \Delta_I f$$

- **Especially when estimating the norm of a linear operator T acting on f**

$$|\langle Tf, g \rangle| \leq C \|f\| \|g\| \quad ?$$

$$\langle Tf, g \rangle = \sum_{I,J} \langle T \Delta_I^\mu f, \Delta_J^\nu g \rangle$$



Key question: bounds for the Hilbert transform

$$Hf(x) = \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{f(y) dy}{x - y}$$

Classical Fourier analysis: $\widehat{Hf}(\xi) = -i \operatorname{sign}(\xi) \hat{f}(\xi)$

$$\begin{aligned} \|Hf\|_{L^2} &= \left(\int_{-\infty}^{\infty} |Hf(x)|^2 dx \right)^{1/2} \\ &= \|\widehat{Hf}\|_{L^2} = \|\hat{f}\|_{L^2} = \|f\|_{L^2} \end{aligned}$$

But for the *weighted* Hilbert transform...

$$H(f d\mu)(x) = \int_{-\infty}^{\infty} \frac{f(y) d\mu(y)}{x - y} \quad \left| \int H(f d\mu)g d\nu \right| \leq C \|f\|_{L^2(\mu)} \|g\|_{L^2(\nu)}?$$

...need dyadic analysis!



Dyadic analysis of the weighted Hilbert transform

$$\int H(\Delta_I^\mu f \, d\mu) \Delta_J^\nu g \, d\nu = \iint_{I \times J} \frac{1}{x - y} \Delta_I^\mu f(y) \Delta_J^\nu g(x) \, d\mu(y) \, d\nu(x)$$

Difficulties:

- Division by zero if $x = y$
- Discontinuity if y lies at centre or boundary of I
- Or if x lies at centre or boundary of J

Worst case: all at once, say $x \approx y \approx \partial I \approx \partial J$ ($\partial =$ boundary)

In particular: $|I| \ll |J|$ $\text{dist}(I, \partial J) \ll |J|$ – ‘bad’ case!

But this should be ‘rare’ – more ‘likely’ to be in the interior than the boundary.



Enter probability: random dyadic cubes

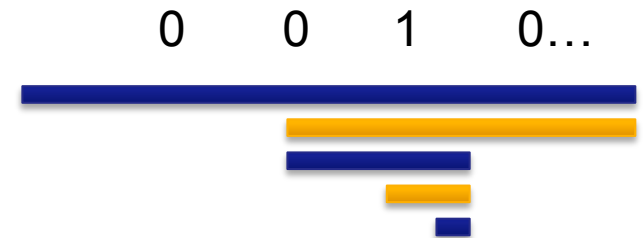
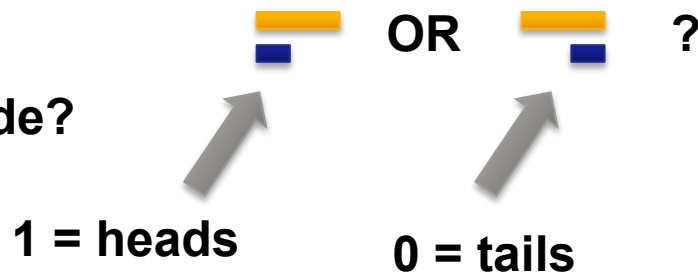
(F. Nazarov, S. Treil, A. Volberg)

- Start from a given finest scale of dyadic intervals, coarser scales to be chosen



- For a representative of these intervals, need to decide if it will be the left or right half of its dyadic 'parent'

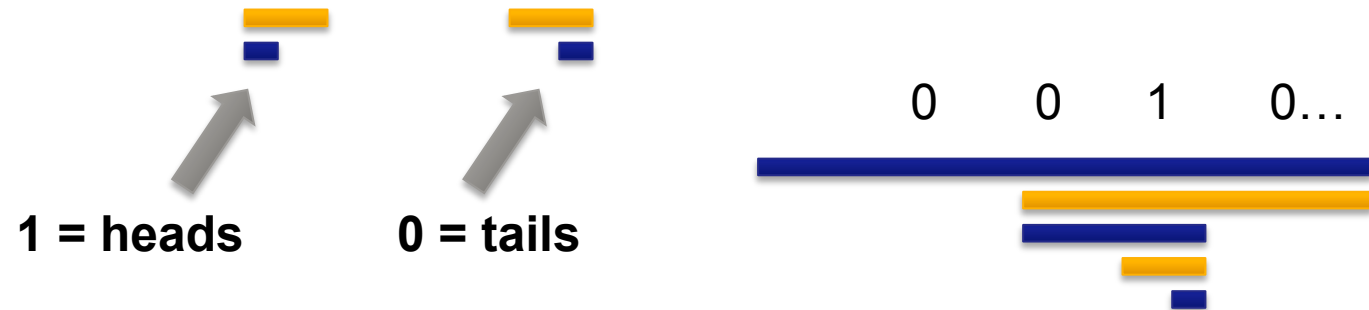
- How to decide?



Repeat for all coarser levels.



Random dyadic cubes, continued



- **Probability space: an infinite product of coin tosses**

$$\Omega = \{0, 1\}^{\mathbb{Z}} \ni \omega = (\omega_j)_{j \in \mathbb{Z}}$$

- **Random intervals = random translates of standard intervals**

$$I \dot{+} \omega := I + \sum_{\substack{j \in \mathbb{Z} \\ 2^{-j} < |I|}} 2^{-j} \omega_j \sim I + u \quad u \sim \text{Unif}[0, |I|]$$

- **A random translate of I is 'bad' if for some much bigger J**

$$2^r |I| < |J| \quad \text{dist}(I \dot{+} \omega, \partial(J \dot{+} \omega)) \leq |I|^\gamma |J|^{1-\gamma} = \left(\frac{|I|}{|J|} \right)^\gamma |J|$$

1 = heads

- **Rigorous probabilistic bound** $\mathbb{P}(I \dot{+} \omega \text{ is bad}) \leq c_\gamma 2^{-r\gamma}$



Random cubes and bounds for operators

- **Splitting into good and bad parts:**

$$f = \sum_{I: I \dot{+} \omega \text{ is good}} \Delta_{I \dot{+} \omega}^\mu f + \sum_{I: I \dot{+} \omega \text{ is bad}} \Delta_{I \dot{+} \omega}^\mu f =: f_{\text{good}} + f_{\text{bad}}$$

$$\langle Tf, g \rangle = \langle Tf_{\text{good}}, g_{\text{good}} \rangle + \langle Tf_{\text{bad}}, g_{\text{good}} \rangle + \langle Tf, g_{\text{bad}} \rangle$$

- **Good part: ‘direct’ estimates, valid for any ω**

$$|\langle Tf_{\text{good}}, g_{\text{good}} \rangle| \leq C \|f\| \|g\|$$

- **Bad part: ‘indirect’ estimate, valid only on average**

$$\mathbb{E} |\langle Tf_{\text{bad}}, g_{\text{good}} \rangle + \langle Tf, g_{\text{bad}} \rangle| \leq \epsilon \|T\| \|f\| \|g\|$$

if $|\langle Tf, g \rangle| \leq \|T\| \|f\| \|g\|$

- **Synthesis:**

$$\|T\| \leq C + \epsilon \|T\| \quad \Rightarrow \quad \|T\| \leq \frac{C}{1 - \epsilon}$$



Achievements based on random dyadic intervals:

Nazarov-Treil-Volberg's characterization (~ 2000) of the measures for which:

$$|\langle H(f \, d\mu), g \, d\mu \rangle| \leq C \|f\|_{L^2(\mu)} \|g\|_{L^2(\mu)}$$

In 2D (the complex plane), this led to the solution of Painlevé's problem:

Which planar sets E have the removability property that

if a function f is bounded and analytic outside E , then it has an extension to E that remains bounded and analytic?



Solution to Painlevé's problem by X. Tolsa (2003):

A compact set E is non-removable for bounded analytic functions

if and only if

it supports a positive Radon measure with linear growth and finite curvature

$$\mu(D(x, r)) \leq Cr \quad \iiint_{\mathbb{C} \times \mathbb{C} \times \mathbb{C}} \frac{d\mu(x) d\mu(y) d\mu(z)}{R(x, y, z)^2} < \infty$$



Solution to the two-weight problem by M. Lacey (2013):

**For a pair of measures (μ, ν)
(with no common point masses)**

$$|\langle H(f \, d\mu), g \, d\nu \rangle| \leq C \|f\|_{L^2(\mu)} \|g\|_{L^2(\nu)}$$

if and only if

it holds whenever either $f = 1_I$ or $g = 1_I$

(Sawyer-type ‘testing condition’) and

$$\sup_{(x,t) \in \mathbb{R}_+^2} P(\mu)(x,t) \cdot P(\nu)(x,t) < \infty$$

(the ‘Poisson A_2 condition’).



Beyond Euclidean spaces:

- Principle of ‘dyadic’ decomposition is very general

Metric measure space (X, d, μ)

- a set X equipped with
- ‘distance’ d and
- ‘measure’ μ (‘mass’ / ‘volume’ / etc.)

Example: X = ‘the Internet’ (= all devices connected to it)

$d(x, y)$ = time it takes to transfer 1 Mb of data from x to y

$\mu(E)$ = total data storage capacity of all devices $x \in E$



Metric space ‘dyadic cubes’ of M. Christ (1990)

- **Built from ‘centres’ and ‘parent-child’ relation between ‘cubes’ of different generation**
- **Randomization by T.H. & H. Martikainen (2012): for every cube, pick a child (randomly), let its centre be the centre of the new cube**

Abstract extensions of many results above



Smoother wavelets

$$f = \sum_{j,k} \psi_{j,k} \langle \psi_{j,k}, f \rangle$$

$\psi_{j,k}$ 'smoothly' localized around $I_{j,k}$

- **Often preferred over Haar in analysis on the Euclidean space with Lebesgue measure**

- **What about more general spaces?**



General construction of wavelets from a multiresolution analysis (Y. Meyer)

$$\{0\} \subseteq \dots \subseteq V_{j-1} \subseteq V_j \subseteq V_{j+1} \subseteq \dots \subseteq L^2(\mu)$$

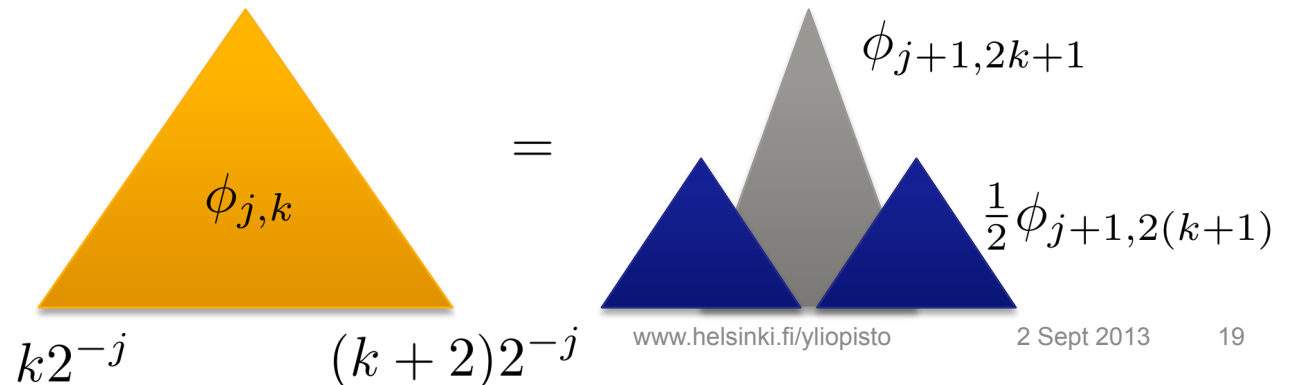
$$V_j = \text{span}\{\phi_{j,k} : k \in \mathcal{K}_j\}$$

Then we can find wavelets with same regularity

$$V_{j+1} = V_j \oplus W_j \quad W_j = \text{span}\{\psi_{j,k} : k \in \mathcal{K}'_j\}$$

Example: piecewise linear splines

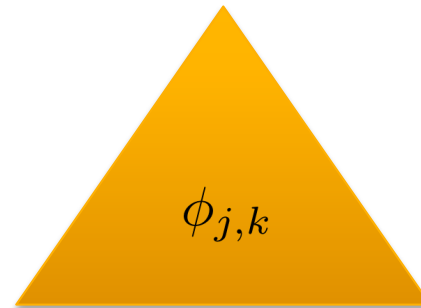
$$V_j = \{f \text{ linear on each } [\frac{k}{2^j}, \frac{k+1}{2^j}], k \in \mathbb{Z}\} = \text{span}\{\phi_{j,k} : k \in \mathbb{Z}\}$$



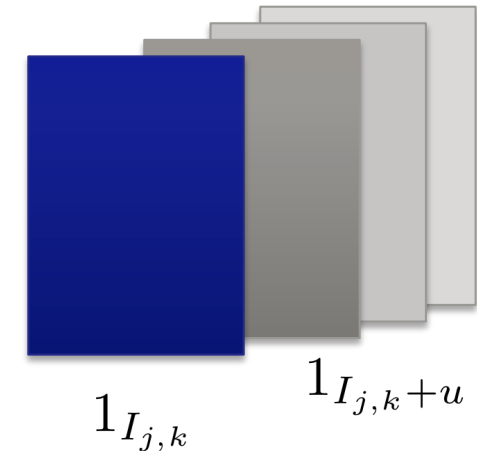


Interpretation via probability:

$$\phi_{j,k}(x) = 2^j \int_0^{2^{-j}} 1_{I_{j,k+u}}(x) du = \mathbb{E}_\omega 1_{I_{j,k} + \omega}$$



= average of



**Using the abstract random dyadic cubes:
first continuous splines & wavelets in metric
measure spaces by P. Auscher & T.H. (2013)**

$$\phi_{j,k}(x) := \mathbb{E}_\omega 1_{I_{j,k}(\omega)}$$



Thank you!

