

Exact recovery from ‘projections’ onto polynomial spaces over compact manifolds

Shai Dekel (GE Global Research & Tel-Aviv University)

Joint work with: Tamir Bendory and Arie Feuer (Technion).

Case I: Trigonometric polynomials

- Spike train $f(t) = \sum_{m=1}^M a_m \delta_{t_m}(t)$, δ_x Dirac,

$a_m \in \mathbb{C}$ coefficients, $-\pi \leq t_1 < t_2 < \dots < t_M < \pi$ knots.

- $\langle f, g \rangle = \sum_m a_m g(t_m)$, $\forall g \in C[-\pi, \pi]$.

- f in dual space of Borel measures $\mathcal{M}([-\pi, \pi])$.

- Total variation of a complex measure over a compact $A \subset \mathbb{R}^n$

$$\|\mu\|_{TV} = |\mu|(A) := \sup \sum_k |\mu(A_k)|, \quad A = \bigcup_k A_k, \text{ interior disjoint finite}$$

collection.

• For a spike train $f(t) = \sum_m a_m \delta_{t_m}(t)$, $\|f\|_{TV} = \sum_m |a_m|$.

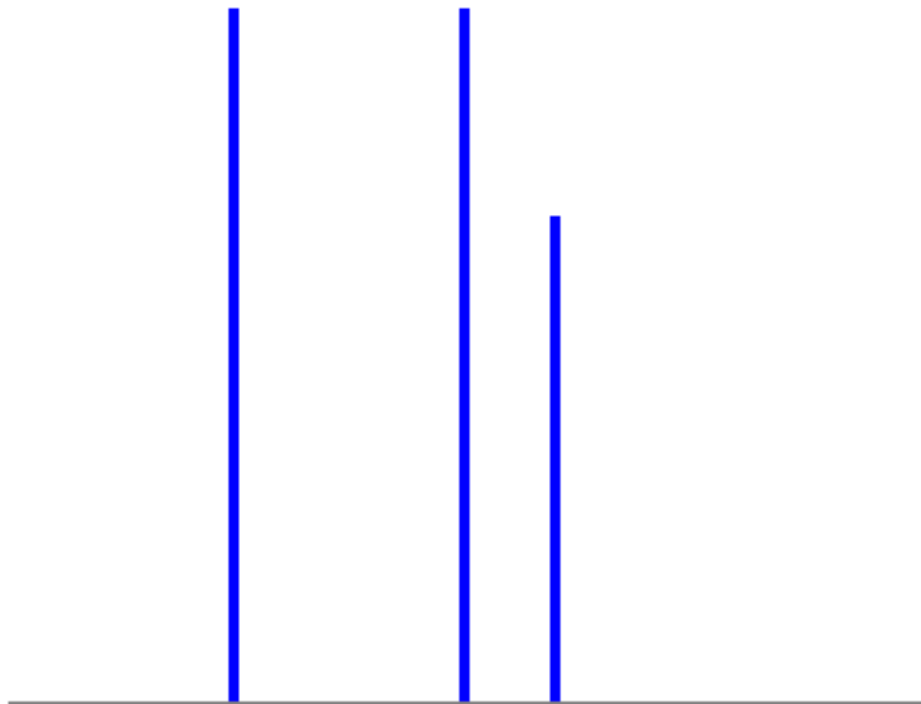
• **Input:** For some degree N , Fourier ‘coefficients’:

$$y_k := \langle f, e^{ik\cdot} \rangle = \frac{1}{2\pi} \sum_m a_m e^{-ikt_m}, \quad -N \leq k \leq N.$$

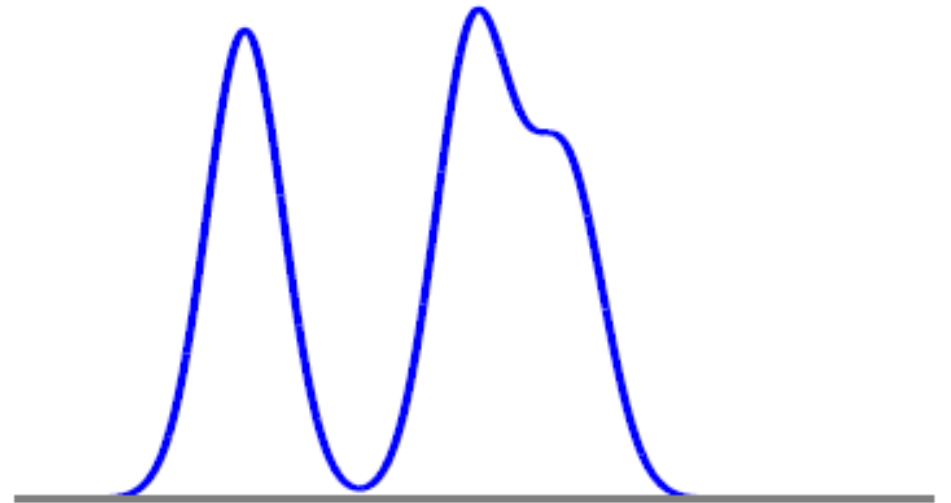
Note: number of spikes and the coefficients are unknown.

• **Goal:** Recover f exactly from $\{y_k\}_{k=-N}^N$.

• E Candés & C. Fernandez-Granda, Towards a Mathematical Theory of Super-Resolution, 2012.



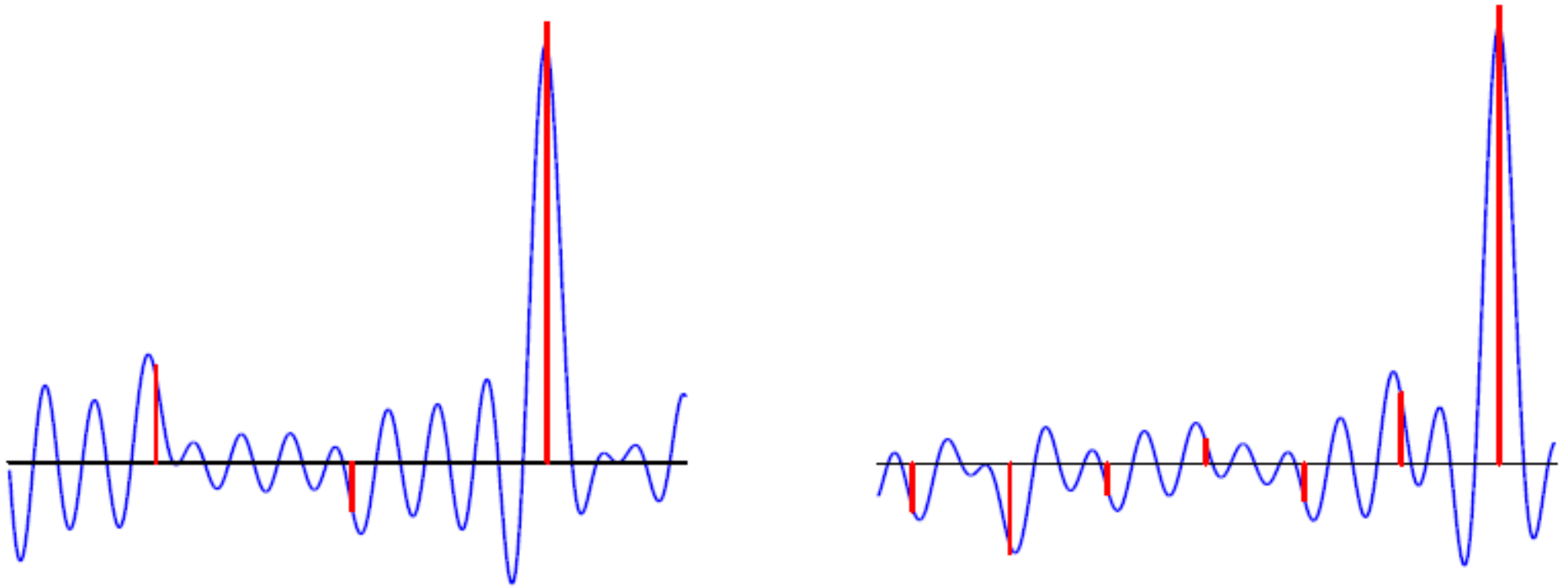
(a)



(b)

Spikes and their lower-resolution ‘projection’

(*) E Candés & C. Fernandez-Granda, Towards a Mathematical Theory of Super-Resolution, 2012.



Two different trains with qualitatively similar ‘projections’

(*) E Candés & C. Fernandez-Granda, Towards a Mathematical Theory of Super-Resolution, 2012.

- **Separation condition [CF]** Assume that for $N \geq 128$

$$\Delta(T) := \min_{j,k} |t_j - t_k| \geq \frac{4\pi}{N} \quad (\text{Cyclic distance}).$$

Theorem [CF 2012] If the knots of a spike train f satisfy the separation condition and $y_k = \langle f, e^{ik\cdot} \rangle$, $-N \leq k \leq N$, are given, then f is the *unique complex measure* solving

$$\min_{\tilde{f} \in \mathcal{M}[-\pi, \pi]} \|\tilde{f}\|_{TV}, \quad \text{s.t. } \langle \tilde{f}, e^{ik\cdot} \rangle = y_k, \quad -N \leq k \leq N.$$

- CF also analyzed the ‘applied setting’:
 - (i) Resilience to noise,
 - (ii) Stable recovery algorithm.

The ‘dual interpolating polynomial’

Theorem 7. *Let $f = \sum_m c_m \delta_{x_m}$ where $X := \{x_m\} \subseteq A$, and $A \subset \mathbb{R}^n$ is compact. Let Θ_D be a linear space of continuous functions of dimension $D+1$ in A . For any basis $\{\theta_k\}_{k=0}^D$, of Θ_D , let $y_k = \langle f, \theta_k \rangle$ for all $0 \leq k \leq D$. If for any set $\{u_m\}$, $u_m \in \mathbb{C}$, with $|u_m| = 1$, there exists $q \in \Theta_D$ such that*

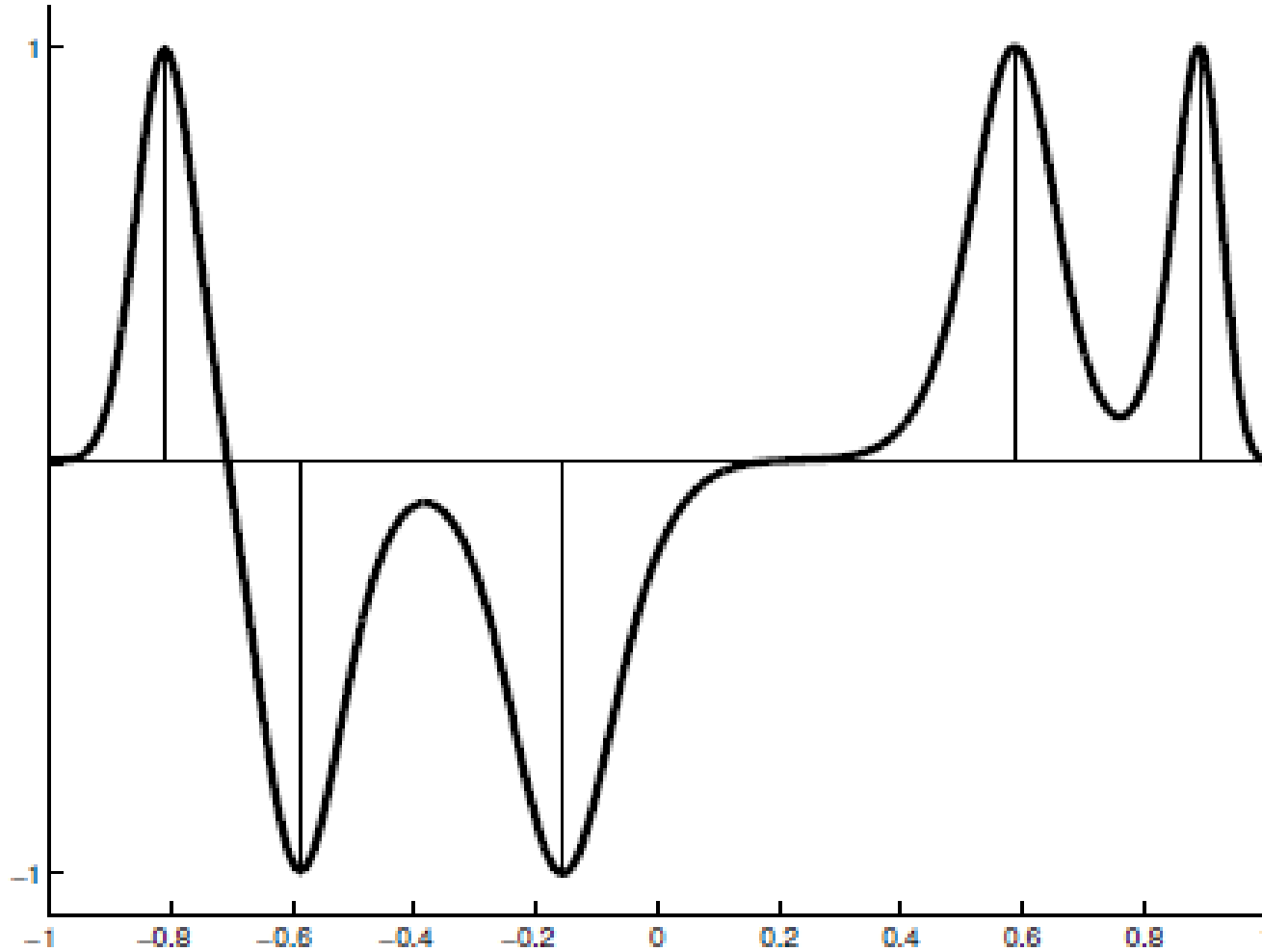
$$q(x_m) = u_m, \quad \forall x_m \in X, \quad (2.1)$$

$$|q(x)| < 1, \quad \forall x \in A \setminus X, \quad (2.2)$$

then f is the unique complex Borel measure satisfying

$$\min_{g \in \mathcal{M}(A)} \|g\|_{TV} \quad \text{subject to} \quad y_k = \langle g, \theta_k \rangle, \quad 0 \leq k \leq D. \quad (2.3)$$

Constructions of interpolating polynomials (with certain properties) \Rightarrow Exact recovery of spike trains through TV-minimization.



An interpolating **algebraic polynomial** with values $u_m \in \{-1, 1\}$ at specified knots

Non-negative signals

- Assume the unknown spike-train is of form

$$f(t) = \sum_{m=1}^M a_m \delta_{t_m}(t), \quad a_m > 0, \quad t \in [-\pi, \pi].$$

Theorem If $M \leq N$, then f is the unique minimizer over all non-negative real measures of

$$\min_{\tilde{f} \in \mathcal{M}_{\geq 0}[-\pi, \pi]} \|\tilde{f}\|_{TV}, \quad \text{s.t. } \langle \tilde{f}, e^{ik\cdot} \rangle = y_k, \quad -N \leq k \leq N.$$

Remark: Discrete ‘non-negative’ version known from Donoho-Tanner (2005). Very different arguments.

Here, we use a simple version of the ‘duality’ theorem for non-negative measures (the polynomial needs to be real).

The construction of q :

$$q(t) := 1 - 2^{-M-1} \prod_{m=1}^M (1 - \cos(t - t_m))$$

q is a real trigonometric polynomial that satisfies

- (i) q is a polynomial of degree $\leq N$.
- (ii) $q(t_m) = 1$, for $1 \leq m \leq M$.
- (iii) $0 < q(t) < 1$, for $t \neq t_m$, $1 \leq m \leq M$.

Case II: Algebraic polynomials

- Spike train $f(x) = \sum_m a_m \delta_{x_m}(x)$, δ_t Dirac,

$a_m \in \mathbb{C}$ coefficients, $-1 \leq x_1 < x_2 < \dots < x_M < 1$ knots.

- **Input:** For some degree N and polynomial basis $\{P_k\}$ of V_N :

$$y_k := \langle f, P_k \rangle = \sum_m a_m P_k(x_m), \quad 0 \leq k \leq N.$$

Recall: number of spikes and the coefficients are unknown.

- **Goal:** Recover f exactly from $\{y_k\}_{k=-N}^N$.

- **Separation condition** Assume for $N \geq 128$, that the knots satisfy

$$\frac{x_{k+1} - x_k}{\sqrt{1 - \bar{x}^2}} \geq \frac{4\pi}{N}, \quad \bar{x} := \arg \min_{z \in [x_k, x_{k+1}]} |z|,$$

(+ another technical condition)

- Condition aligns with the classical metric over the interval. We can allow closer knots near the endpoints.

Theorem If the knots of a spike train f satisfy the (algebraic) separation condition and $y_k = \langle f, P_k \rangle$, $0 \leq k \leq N$, are given, then f is the *unique complex measure* solving

$$\min_{\tilde{f} \in \mathcal{M}[-1,1]} \|\tilde{f}\|_{TV}, \quad \text{s.t. } \langle \tilde{f}, P_k \rangle = y_k, \quad 0 \leq k \leq N.$$

Algebraic polynomials over $[-1,1]^2$

- Spike train $f(x) = \sum_m a_m \delta_{x_m}(x)$, δ_t Dirac,

$$a_m \in \mathbb{R} \text{ coefficients, } x_m = (x_m(1), x_m(2)) \in (-1,1)^2 .$$

- **2D separation condition** Assume for $N \geq 128$, that

$$\min_{j,k} \left\{ \frac{|x_k(1) - x_j(1)|}{\sqrt{1 - \bar{x}_1^2}}, \frac{|x_k(2) - x_j(2)|}{\sqrt{1 - \bar{x}_2^2}} \right\} \geq \frac{4\pi}{N},$$

(+ technical condition).

Theorem If the knots of a 2D spike train f satisfy the 2D separation condition and $y_k = \langle f, P_k \rangle$, $k = (k_1, k_2)$, $0 \leq k_1, k_2 \leq N$, are given, then f is the *unique real measure* solving

$$\min_{\tilde{f} \in \mathcal{M}[-1,1]^2} \|\tilde{f}\|_{TV}, \quad \text{s.t. } \langle \tilde{f}, P_k \rangle = y_k .$$

Spline case

Assume the unknown f is a piecewise constant (order $r = 1$)

$$f(t) = c_0 \mathbf{1}_{[-1, t_1)} + \sum_{m=1}^{M-1} c_m \mathbf{1}_{[t_m, t_{m+1})}(t) + c_M \mathbf{1}_{[t_M, 1]}(t),$$

With known:

- (i) boundary conditions $f(-1) = c_0$, $f(1) = c_M$ (M unknown!).
- (ii) $y_k = \langle f, P_k \rangle$, $\{P_k\}$ some polynomial basis of V_N .

The distributional derivative is a spike train

$$f'(t) = \sum_{m=1}^{M-1} (c_m - c_{m-1}) \delta_{t_m}(t).$$

What are $\{\langle f', P_k \rangle\} = ?$

- Let $\{\alpha_{k,n}\}$ coefficients such that

$$P'_k = \sum_n \alpha_{k,n} P_n, \quad \forall k .$$

- Then with $y_k = \langle f, P_k \rangle$, integration by parts gives

$$\langle f', P_k \rangle = f(1)P_k(1) - f(-1)P_k(-1) - \sum_n \alpha_{k,n} y_n .$$

- Exact recovery of the spike train f' yields exact recovery of the piecewise constant f .
- Generalization to spline of arbitrary order r (degree $r-1$):
 - Assume boundary conditions $f^{(j)}(-1), f^{(j)}(1)$, $j = 0, \dots, r-1$, are known.
 - Recover (via recursion) f from the spike train $f^{(r)}$.

Case III: Spherical Harmonics on \mathbb{S}^{d-1}

- $\mathbb{Y}_n(\mathbb{R}^d)$ Homogeneous Harmonic polynomials of degree n .
- $\mathbb{Y}_n^d := \mathbb{Y}_n(\mathbb{R}^d)|_{\mathbb{S}^{d-1}}$ Spherical Harmonics of degree n in d dimensions.
- Generalization of trigonometric polynomials.
- Let $\{P_{n,k}\}$, $n \leq N$, $1 \leq k \leq Z_{n,d}$ be a basis for spherical harmonics of degree N

$$Z_{n,d} = \frac{(2n+d-2)(n+d-3)!}{n!(d-2)!}$$

$$f(\xi) = \sum_m a_m \delta_{\xi_m}(\xi), \quad a_m \in \mathbb{R}, \quad \Xi = \{\xi_m\}, \quad \xi_m \in \mathbb{S}^2 \subset \mathbb{R}^3.$$

- **Sphere separation condition** For a fixed constant $\nu > 1$, assume the points $\Xi = \{\xi_m\}$ satisfy for (sufficiently large) N ,

$$\min_{\xi_j, \xi_k \in \Xi} \arccos(\xi_j \cdot \xi_k) \geq \frac{\nu}{N}.$$

Theorem If the set $\Xi = \{\xi_m\}$ satisfies the separation condition for sufficiently large N and $y_{n,k} = \langle f, P_{n,k} \rangle$, $n \leq N$, are given, then f is the *unique real measure* solving

$$\min_{\tilde{f} \in \mathcal{M}(\mathbb{S}^2)} \|\tilde{f}\|_{TV}, \quad \text{s.t.} \quad \langle \tilde{f}, P_{n,k} \rangle = y_{n,k}.$$

So what's under the hood? A localization principle!



Localization principle

- Recall that we need to construct q , such that
 - (i) $q(t_m) = u_m$, t_m knots, $u_m \in \mathbb{C}$, $|u_m| = 1$ prescribed.
 - (ii) $|q(t)| < 1$, $t \neq t_m$.
- Proof relies on finding **well-localized polynomial kernel** $K_N(x, y)$ of the given degree N .
- Trigonometric polynomials

$$q(t) = \sum_m (\alpha_m K(t - t_m) + \beta_m K'(t - t_m)),$$

$K_N(x, y) = K(x - y)$, the Jackson kernel, $\{\alpha_m\}, \{\beta_m\}$ selected to satisfy conditions ($q(t_m) = u_m, \dots$).

- Algebraic polynomials
 - **Well localized kernel exists**, but not translation invariant!
 - Direct proof exists...but easier to reduce the problem to the trigonometric case.
- Spherical harmonics
 - Rotation invariant **well-localized(!)** kernel,

$$K_N(\xi_1 \cdot \xi_2) = \kappa_N \sum_{n=0}^{\infty} \varphi\left(\frac{n}{N}\right) \mathcal{P}_n(\xi_1 \cdot \xi_2),$$

$\mathcal{P}_n(\xi_1 \cdot \xi_2)$ - ortho-projection onto harmonics of degree n ,

$$\varphi \in C^\infty(\mathbb{R}_+) , \varphi(t) = \begin{cases} 1 & t \leq 1/2, \\ 0 \leq \varphi(t) \leq 1 & 1/2 \leq t < 1, \\ 0 & \text{else.} \end{cases}$$

$$|K_N(\xi_1 \cdot \xi_2)| \leq \frac{c_k}{(1 + N \arccos(\xi_1 \cdot \xi_2))^k}, \quad \xi_1, \xi_2 \in \mathbb{S}^{d-1}.$$

- Lie algebra structure - rotational derivatives.
- With ν the constant from the separation condition, we prove estimates of the type

$$\left\| I - \left(K_N(\xi_j \cdot \xi_k) \right)_{j,k} \right\|_{\infty} \leq \frac{c_k}{\nu^{k-1}}.$$