Uncertainty principles and sparse approximation

In this lecture, we will consider the special case where the dictionary Φ is composed of a pair of orthobases. We will see that our ability to find a sparse approximation to a signal (when one exists) hinges on the existence of an *uncertainty principle* between the two bases. Essentially, this uncertainty principle ensures that it would take many elements from one of the bases to represent the space spanned by a small number of elements from the other basis — signals that are concentrated in one of the bases must be spread out in the other. We will see how this type of relationship ensures that sparse combinations of the dictionary elements can only be written one way with a small number of terms.

Spites & Sinusoids
Let's consider the special case where our dictionary
is the union of two orthobases: the "spite" basis
(identity) and the Fourier basis (sinusoids).

$$\vec{P} = \begin{bmatrix} I & F \end{bmatrix} = n \times 2n$$
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 $\vec{P} = (n + 1)(e - 1)/n$
 $\vec{P} = \vec{P} + \vec{P} +$

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It is trivially true that

$$I_{max}(M^{*}M) \leq Trace(M^{*}M) = \sum_{\substack{W \in T}} (M^{*}M)_{w,w}$$

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For each we
$$\Lambda$$

 $(M^{*}M)_{\omega,\omega} = \frac{1}{N} \sum_{f \in T} e^{-j\frac{2\pi\omega T}{N}j\frac{2\pi\omega T}{N}}$
 $= \frac{1T!}{N}$
 $\Rightarrow True(M^{*}M) = \frac{12!\cdot 1T!}{N}$

$$\begin{split} f(t) &= \sum_{\substack{x \in T}} \alpha'_{x} \delta(t-x) + \sum_{\substack{u \in T}} \alpha'_{u} \prod_{\substack{x \in T}} e^{j2\pi u t/n} \\ f &= \overline{\mathbb{P}} \mathbb{K} = \left[\prod F_{2}^{j} \left[\frac{\mu^{j}}{\mu^{2}} \right] & \text{where } \mu^{j} \text{ is nonzero on } \mathcal{I}. \\ \text{Suppose that } \mathbb{K} \text{ has fewer than } \sqrt{n} \text{ non-zero } \\ \text{terms:} & |T| + |\mathcal{I}_{2}| < \sqrt{n} \\ \text{is there another way to write f with a } \\ \text{comparable number of spikes and sinusoids ?} \\ \frac{N_{0}}{\frac{1}{\mu^{2}}} \\ \text{Why?} \quad \text{Suppose there were another way, that is } \\ \text{suppose } \exists \beta \neq \mathbb{K} \text{ such that } \\ \overline{\mathbb{F}} \beta = \overline{\mathbb{P}} \mathbb{K} = f \\ \text{and} \\ & |\text{supp } \beta^{j}| (+|\text{supp } \beta^{2}| \leq |T| + |\mathcal{N}| < \sqrt{n} \\ \text{Set } h = \mathbb{K} - \beta, \text{ note that } h \in Mull(\overline{\mathbb{P}}) \\ \overline{\mathbb{F}} h = \overline{\mathbb{P}} \mathbb{K} - \overline{\mathbb{F}} \beta = f - f = O \end{split}$$

Also note that

$$|supp h| \leq |T| + |J_{L}| + |supp \beta'| + |supp \beta^{2}|$$

$$< 2\sqrt{n}$$
But what do vectors in Nall(\overline{E}) lock like?

$$\overline{E}h = 0 \Rightarrow [T F] [\frac{h'}{n^{2}}] = 0$$

$$\Rightarrow h' + Fh^{2} = 0 \Rightarrow h^{2} = -F^{*}h'$$

$$\Rightarrow hey have the form
$$\begin{bmatrix} h'\\ -h'\end{bmatrix}$$
By the UP, any such vector must have

$$support at least 2\sqrt{n}, which contradicts$$

$$=he above.$$

$$\Rightarrow K is the only decomposition which uses
fever than \sqrt{n} terms.$$$$

As a direct result, if we observe f and there exits
an expansion of f using fewer them doit terms, solving
min liplleo
c.t.
$$\Xi\beta$$
=f
will find it (and the solution is unique).

 $l_1 \notin Sparsest decomposition$
It is also possible to develop an uncertainty
principle for concentration in the l_1 -norm
(see Donoho \notin thuo, 2001). Using this, it
can be shown that if f has a
decomposition & with
 $|supp K| = |T| + |M| \lesssim .9 \sqrt{n}$
the solution to
min liplle,
s.t. $\Xi\beta$ =f
will be unique and equal to K

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Pairs of Bases
The min results can be generalized past "spikes + Forrier".
Given a dictionary which is composed of Two orthobones

$$\overline{\Psi} = \begin{bmatrix} \Psi_1 & \Psi_2 \end{bmatrix}$$

define

Note that

$$| \leq M \leq \sqrt{n}$$
Just as with spikes + Fourier, we can prove a UP:
Thui (Elad & Bruckstein, '02)
Let $f \in C^{\circ}$ be given, and set
 $\beta_1 = \Psi_1^* f$ (Ψ_1 -transform)
 $\beta_2 = \Psi_2^* f$ (Ψ_2 -transform)
Then
Isupp $\beta_1 |\cdot| | supp |\beta_2 | \geq n/4^2$
 $\Rightarrow | | supp |\beta_1 |+| | supp |\beta_2 | \geq 2\sqrt{n}/4$

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This leads directly to the uniqueness result:
This suppose
$$f = \overline{E}\kappa$$
, where
Is upp $\kappa I < \sqrt{n}/M$
Then κ is the only decomposition of f that
uses fever than \sqrt{n}/M terms, and is the unique
solution TO
min $H\beta II_{R}$
 $s.t. \overline{E}\beta = f$
An L_1 result exists as well:
Then $(\overline{E} + \beta + 02)$:
If $f = \overline{E}\kappa$ with
Isopp $\kappa I < \frac{1}{2}(1 + \sqrt{n}/M)$ (~ \sqrt{n}/M)
then κ is the unique solution to the convex
problem
min II βII_{R}
 $s.t. \overline{E}\beta = f$

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=> "untangling" the contributions from the two different bases requires M TO be small i.e. The bases must be "unsimilar"

with

$$|T| + |D| \leq Const. \frac{n}{\sqrt{\log n}}$$
 (Const ~ 1/8)
Then with high probability, it is impossible TO
find a signal fell supported on T such that
f is supported on Λ .
(In fur, it is impossible that even half of the energy
of f can be supported on Λ .)

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This theorem says that a UP holds for "generic"
(i.e. minus special cases like the Dirac camb) signals
UPTO sparsity ~ Nation (conjunc to ~Nat).
There are also associated to
$$\pm l_1$$
 recovery results:
Them: Let $K = \begin{bmatrix} 0^{1/2} \\ k^2 \end{bmatrix}$ be a space decomposition of
 $f = \overline{\Phi}K$, $\overline{\Phi} = [I F]$, with X' supported on T and
 X^2 supported on \mathcal{N} , where $T \neq \mathcal{N}$ are chosen
as above. Then for the "vest majority" of such
 X , K will be the unique solution to
min HpBleo
s.t. $\overline{\Phi}\beta = f$
where $f = \overline{\Phi}K$.
The l_1 result requires a little bit more sparsity
Them: Chose T, \mathcal{R} uniformly at random with
 $|T| + |\mathcal{N}| \leq Const. \frac{N}{\log N}$
Then for the voot majority of K supported on TUM;
 K is the unique solution to
min lipble.
 $St. \overline{\Phi}\beta = f$

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Exensions TO general pairs of bases hold as
Vell- replace n with n/42 everywhere above.
Leaving out "special cases", l, can effectively
recover decompositions with spasity on the order
of ~ n/logn (not two far from n).
Bounds like these can also be found for
general dictributaries \$ (not necessarily unions-
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of
$$\bot$$
-bases) - the Key parameter in this
Case is
 $M = \max_{\substack{X \neq X \\ Y \neq X}} | K P_X, P_X > |$
(See work by Elad & Donoho '03, # Tropp '06).