

# SALEM SETS AVOIDING ROUGH CONFIGURATIONS

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Recall that a set  $X \subset \mathbf{R}^d$  is a *Salem set* of dimension  $s$  if it has Hausdorff dimension  $s$ , and for every  $\varepsilon > 0$ , there exists a probability measure  $\mu_\varepsilon$  supported on  $X$  such that for all  $\xi \in \mathbf{R}^d$ ,

$$\sup_{\xi \in \mathbf{R}^d} |\xi|^{s-\varepsilon} |\widehat{\mu_\varepsilon}(\xi)| < \infty.$$

Our goal in these notes is to obtain, for each set  $Z \subset \mathbf{R}^{dn}$  with Minkowski dimension  $s$ , a Salem set  $X \subset \mathbf{R}^d$  with dimension

$$\frac{nd - s}{s},$$

such that for each set of  $n$  distinct elements  $x_1, \dots, x_n \in X$ ,  $(x_1, \dots, x_n) \notin Z$ . We hope that we can rely on the random selection approach of our paper on rough configurations to obtain such a result.

## 1. ORLICZ NORM

We define a convex function  $\psi_2 : [0, \infty) \rightarrow [0, \infty)$  by  $\psi_2(t) = e^{t^2} - 1$ , and define a corresponding Orlicz norm on the family of scalar valued random variables  $X$  over a probability space by setting

$$\|X\|_{\psi_2(L)} = \inf \{A \in (0, \infty) : \mathbf{E}(\psi_2(|X|/A)) \leq 1\}.$$

The family of random variables  $\psi_2(L)$  are known as *subgaussian random variables*. Here are some important properties:

- (Gaussian Tails): There exists a universal constant  $c > 0$  such that for any random variable  $X$ ,  $\|X\|_{\psi_2(L)} \leq A$  if and only if for each  $t \geq 0$ ,

$$\mathbf{P}(|X| \geq t) \leq 2 \exp(-ct^2/A^2).$$

- (Bounded Variables are Subgaussian): For any random  $X$ ,

$$\|X\|_{\psi_2(L)} \lesssim \|X\|_{L^\infty}.$$

- (Centering) For any random variable  $X$ ,

$$\|X - \mathbf{E}(X)\|_{\psi_2(L)} \lesssim \|X\|_{\psi_2(L)}.$$

- (Union Bound) If  $X_1, \dots, X_N$  are random variables, then

$$\|X_1 + \dots + X_N\|_{\psi_2(L)} \leq \|X_1\|_{\psi_2(L)} + \dots + \|X_N\|_{\psi_2(L)}.$$

- (Hoeffding's Inequality): If  $X_1, \dots, X_N$  are *independent* random variables, then

$$\|X_1 + \dots + X_N\|_{\psi_2(L)} \lesssim \left( \|X_1\|_{\psi_2(L)}^2 + \dots + \|X_N\|_{\psi_2(L)}^2 \right)^{1/2}.$$

The Orlicz norm is a convenient notation to summarize calculations of subgaussian concentration inequalities. Roughly speaking, we can think of a random variable  $X$  with  $\|X\|_{\psi_2(L)} \leq A$  as having the large majority of its support in the interval  $[-A, A]$ , and having mass rapidly decaying outside of this interval.

## 2. A FAMILY OF CUBES

We fix sequences of integers  $\{N_m : m \geq 1\}$  and  $\{M_m : m \geq 1\}$ . We then define two sequences of real numbers  $\{l_m : m \geq 0\}$  and  $\{r_m : m \geq 0\}$ , by initially setting  $l_0 = r_0 = 1$ , and then, for each  $m \geq 1$ , setting  $r_m = l_{m-1}/M_m$ , and  $l_m = r_m/N_m$ . For each  $m \geq 0$  and  $d$ , we define two collections of strings

$$\Sigma_m^d = \mathbf{Z}^d \times [M_1]^d \times \dots \times [N_1]^d \times \dots \times [M_m]^d \times [N_m]^d$$

and

$$\Pi_m^d = \mathbf{Z}^d \times [M_1]^d \times [N_1]^d \times \dots \times [N_{m-1}]^d \times [M_m]^d.$$

For each string  $i \in \Sigma_m^d$ , we define a vector  $a_i \in (l_m \mathbf{Z})^d$  by setting

$$a_i = i_0 + \sum_{k=1}^m i_{2k-1} \cdot r_k + i_{2k} \cdot l_k$$

Then each string  $i \in \Sigma_m^d$  can be identified with a sidelength  $l_m$  cube

$$Q_i = \prod_{j=1}^d [a_{ij}, a_{ij} + l_m].$$

centered at  $a_i$ . Similarly, for each string  $i \in \Pi_m^d$ , we define a vector  $a \in (r_m \mathbf{Z})^d$  by setting, for each  $1 \leq j \leq d$ ,

$$a_i = i_0 + \left( \sum_{k=1}^{m-1} i_{2k-1} \cdot r_k + i_{2k} \cdot l_k \right) + i_{2m-1} \cdot r_m,$$

and then define a sidelength  $r_m$  cube

$$R_i = \prod_{j=1}^d [a_{ij}, a_{ij} + r_m].$$

We let  $\mathcal{Q}_m^d = \{Q_i : i \in \Sigma_m^d\}$ , and  $\mathcal{R}_m^d = \{R_i : i \in \Pi_m^d\}$ . Then

- For each  $m$ ,  $\mathcal{Q}_m^d$  and  $\mathcal{R}_m^d$  are covers of  $\mathbf{R}^d$ .
- If  $Q_1, Q_2 \in \bigcup_{m=0}^\infty \mathcal{Q}_m^d$ , then either  $Q_1$  and  $Q_2$  have disjoint interiors, or one cube is contained in the other. Similarly, if  $R_1, R_2 \in \bigcup_{m=1}^\infty \mathcal{R}_m^d$ , then either  $R_1$  and  $R_2$  have disjoint interiors, or one cube is contained in the other.

- For each cube  $Q \in \mathcal{Q}_m$ , there is a unique cube  $Q^* \in \mathcal{R}_m$  with  $Q \subset Q^*$ . We refer to  $Q^*$  as the *parent cube* of  $Q$ . Similarly, if  $R \in \mathcal{R}_m$ , there is a unique cube in  $R^* \in \mathcal{Q}_{m-1}$  with  $R \subset R^*$ , and we refer to  $R^*$  as the *parent cube* of  $R$ .

We say a set  $E \subset \mathbf{R}^d$  is  $\mathcal{Q}_m$  discretized if it is a union of cubes in  $\mathcal{Q}_m^d$ , and we then let  $\mathcal{Q}_m(E) = \{Q \in \mathcal{Q}_m^d : Q \subset E\}$ . Similarly, we say a set  $E \subset \mathbf{R}^d$  is  $\mathcal{R}_m$  discretized if it is a union of cubes in  $\mathcal{R}_m^d$ , and we then let  $\mathcal{R}_m(E) = \{R \in \mathcal{R}_m^d : R \subset E\}$ . We set  $\Sigma_m(E) = \{i \in \Sigma_m^d : Q_i \in \mathcal{Q}_m(E)\}$ , and  $\Pi_m(E) = \{i \in \Pi_m^d : R_i \in \mathcal{R}_m(E)\}$ . We say a cube  $Q_1 \times \cdots \times Q_n \in \mathcal{Q}_m^{dn}$  is *strongly non diagonal* if  $Q_i^* \cap Q_j^* = \emptyset$  for each  $i \neq j$ .

### 3. A FAMILY OF MOLLIFIERS

We now consider a family of mollifiers, which we will use to smooth out the Fourier transform of the measures we study. Begin by choosing a non-negative  $C^\infty$  function  $\psi$  supported on  $[-1, 1]^d$  such that

$$\int_{\mathbf{R}^d} \psi(x) dx = 1, \quad (1)$$

and for each  $x \in \mathbf{R}^d$ ,

$$\sum_{n \in \mathbf{Z}^d} \psi(x + n) = 1. \quad (2)$$

Since  $\psi$  is  $C^\infty$  and compactly supported, then for each  $t \in [0, \infty)$ , we conclude

$$\sup_{\xi \in \mathbf{R}^d} |\xi|^t |\widehat{\psi}(\xi)| < \infty. \quad (3)$$

Now we rescale to obtain a mollifier at each scale of the argument. For each  $k > 0$ , we let

$$\psi_k(x) = l_k^{-d} \psi(l_k \cdot x).$$

Then  $\psi_k$  is supported on  $[-l_k, l_k]^d$ . Equation (1) implies that for each  $x \in \mathbf{R}^d$ ,

$$\int_{\mathbf{R}^d} \psi_k = 1. \quad (4)$$

Equation (2) implies

$$\sum_{n \in \mathbf{Z}^d} \psi(x + l_k \cdot n) = l_k^{-d}. \quad (5)$$

For each  $\xi \in \mathbf{R}^d$ ,  $\widehat{\psi_k}(\xi) = \widehat{\psi}(l_k \xi)$ , and so in particular, (3) implies that for each  $t \in [0, \infty)$ ,

$$|\widehat{\psi_k}(\xi)| \lesssim_t l_k^{-t} |\xi|^{-t}. \quad (6)$$

These properties are sufficient to mollify the functions we consider.

## 4. DISCRETE LEMMA

**Lemma 1.** Fix  $s \in [1, dn)$  and  $\varepsilon \in [0, (n-s)/4)$ . Let  $T \subset [0, 1]^d$  be a non-empty,  $\mathcal{Q}_m$  discretized set, and let  $\mu_T$  be a smooth probability measure compactly supported on  $T$ , together with a constant  $C$  such that for each  $m \in \mathbf{Z}^d$ ,

$$|\widehat{\mu_T}(m)| \leq C|m|^{-s/2}.$$

Let  $B \subset \mathbf{R}^{dn}$  be a non-empty,  $\mathcal{Q}_{m+1}$  discretized set such that

$$\#(\mathcal{Q}_{m+1}(B)) \leq (1/l_{m+1})^{s+\varepsilon}.$$

Then there exists a constant  $C(\mu_T, n, s)$  such that if

$$M_{m+1} \geq C(T, \mu_T, n, d, s), \quad (7)$$

$$N_{m+1} \geq C(T, \mu_T, n, d, s) \cdot M_{m+1}^{\frac{s+\varepsilon}{dn-s-2\varepsilon}}, \quad (8)$$

and

$$N_{m+1} \geq 10^{1/\varepsilon}, \quad (9)$$

then there exists a  $\mathcal{Q}_{m+1}$  discretized set  $S \subset T$  together with a smooth probability measure supported on  $S$  such that

- For any strongly non-diagonal cube

$$Q = Q_1 \times \cdots \times Q_n \in \mathcal{Q}_{m+1}(B),$$

There exists  $i$  such that  $Q_i \notin \mathcal{Q}_{m+1}(S)$ .

- For any  $m \in \mathbf{Z}^d$ ,

$$|\widehat{\mu}(m)| \leq (C + l_{m+1}^{-1})|m|^{-s/2}$$

*Proof.* For each  $i \in \Pi_{m+1}^d$ , let  $j_i$  be a random integer vector chosen from  $[N_{m+1}]^d$ , such that the family  $\{j_i : i \in \Pi_{m+1}^d\}$  is independent. We define a measure  $\nu_S$  by setting, for each  $x \in \mathbf{R}^d$ ,

$$d\nu_S(x) = r_{m+1}^d \sum_{i \in \Pi_{m+1}^d} \psi_{m+1}(x - a_{ij_i}) d\mu_T(x).$$

We then normalize, defining

$$\mu_S = \frac{\nu_S}{\nu_S(\mathbf{R}^d)}.$$

If we set

$$S = \bigcup \{Q \in \mathcal{Q}_{m+1}^d : \mu_S(Q) > 0\},$$

then by definition,  $S$  is a  $\mathcal{Q}_{m+1}$  discretized set,  $\mu_S$  is supported on  $S$ , and  $S \subset T$ . It now suffices to show that with nonzero probability,  $S$  and  $\mu_S$  satisfy the properties guaranteed by the lemma.

For each  $i \in \Pi_{m+1}(T)$ , define a random measure  $\nu_i$  by setting

$$d\nu_i(x) = r_{m+1}^d \psi_{m+1}(x - a_{ij_i}) d\mu_T(x).$$

Then  $\nu_S = \sum_{i \in \Pi_{m+1}(T)} \nu_i$ . Note that if  $j, j' \in [N_{m+1}]^d$ , then

$$|a_{ij} - a_{ij'}| = |j - j'| \cdot l_{m+1} \lesssim_d N_{m+1} l_{m+1} = r_{m+1},$$

which implies

$$\begin{aligned} & \left| r_{m+1}^d \int_{\mathbf{R}^d} \psi_{m+1}(x - a_{ij}) \mu_T(x) - r_{m+1}^d \int_{\mathbf{R}^d} \psi_{m+1}(x - a_{ij'}) \mu_T(x) \right| \\ & \leq r_{m+1}^d \int_{\mathbf{R}^d} \psi_{m+1}(x) |\mu_T(x + a_{ij}) - \mu_T(x + a_{ij'})| \\ & \lesssim_d r_{m+1}^{d+1} \int_{\mathbf{R}^d} \psi_{m+1}(x) \|\nabla \mu_T\|_{L^\infty(\mathbf{R}^d)} = r_{m+1}^{d+1} \|\nabla \mu\|_{L^\infty(\mathbf{R}^d)}. \end{aligned} \quad (10)$$

Thus (10) implies that for each  $i$ ,

$$\|\nu_i(\mathbf{R}^d) - \mathbf{E}(\nu_i(\mathbf{R}^d))\|_{L^\infty} \lesssim_d r_{m+1}^{d+1} \|\nabla \mu\|_{L^\infty(\mathbf{R}^d)}. \quad (11)$$

Furthermore, (5) implies

$$\begin{aligned} \sum_{i \in \Pi_{m+1}^d} \mathbf{E}(\nu_i(\mathbf{R}^d)) &= r_{m+1}^d \sum_{(i,j) \in \Sigma_{m+1}^d} \mathbf{P}(j_i = j) \int_{\mathbf{R}^d} \psi_{m+1}(x - a_{ij}) \mu_T(x) dx \\ &= \frac{r_{m+1}^d}{N_{m+1}^d} \int_{\mathbf{R}^d} \left( \sum_{(i,j) \in \Sigma_{m+1}^d} \psi_{m+1}(x - a_{ij}) \right) \mu_T(x) dx \\ &= \frac{r_{m+1}^d l_{m+1}^{-d}}{N_{m+1}^d} = 1. \end{aligned} \quad (12)$$

For all but  $O_d(r_{m+1}^{-d})$  indices  $i$ ,  $\nu_i = 0$  almost surely. Thus we can apply the triangle inequality together with (11), and (12), we conclude

$$\begin{aligned} \|\nu_S(\mathbf{R}^d) - 1\|_{L^\infty} &= \left\| \sum_{i \in \Pi_{m+1}^d} [\nu_i(\mathbf{R}^d) - \mathbf{E}(\nu_i(\mathbf{R}^d))] \right\|_{L^\infty} \\ &\leq \sum_{i \in \Pi_{m+1}^d} \|\nu_i(\mathbf{R}^d) - \mathbf{E}(\nu_i(\mathbf{R}^d))\|_{L^\infty} \\ &\lesssim_d r_{m+1}^{-d} r_{m+1}^{d+1} \|\nabla \mu\|_{L^\infty(\mathbf{R}^d)} \\ &= r_{m+1} \|\nabla \mu\|_{L^\infty(\mathbf{R}^d)}. \end{aligned} \quad (13)$$

Thus if (7) holds for an appropriately chosen constant depending on  $d$  and  $\|\nabla \mu\|_{L^\infty(\mathbf{R}^d)}$ , we can apply (13) to conclude

$$\|\nu_S(\mathbf{R}^d) - 1\|_{L^\infty} \leq 1/2. \quad (14)$$

Thus normalizing by  $\nu_S(\mathbf{R}^d)$  only introduces a negligible constant.

For each  $i \in \Pi_{m+1}^d$ , let

$$S_i = \bigcup \{Q \in \mathcal{Q}_{m+1}^d : \nu_i(Q) > 0\}.$$

Then  $S = \bigcup_{i \in \Pi_{m+1}^d} S_i$ . Because  $j_i$  is selected uniformly from  $[N_{m+1}]^d$  for each  $i$ , and  $\psi_{m+1}$  is supported on  $[-l_{m+1}, l_{m+1}]^d$ ,

$$S_i \subset \bigcup \{R_{i_0} : R_{i_0} \cap R_i \neq \emptyset\}.$$

For any cube  $Q_{ij} \in \Sigma_{m+1}^d$ , there are  $O_d(1)$  pairs  $(i_0, j_0)$  such that  $Q_{i_0 j_0} \cap Q_{ij} \neq \emptyset$ , and so a union bound gives

$$\mathbf{P}(Q_{ij} \in \mathcal{Q}_{m+1}(S)) \leq \sum_{Q_{i_0 j_0} \cap Q_{ij} \neq \emptyset} \mathbf{P}(j_{i'} = j') \lesssim_d N_{m+1}^{-d}.$$

Without loss of generality, removing cubes from  $B$  if necessary, we may assume all cubes in  $B$  are strongly non-diagonal. Let  $Q = Q_{i_1 j_1} \times \cdots \times Q_{i_n j_n}$  be a strongly non-diagonal cube in  $\mathcal{Q}_{m+1}(B)$ . Since  $Q$  is strongly diagonal, the events  $\{Q_{i_k j_k} \in S\}$  are independent from one another, which implies that

$$\mathbf{P}(Q \in \mathcal{Q}_{m+1}(S^n)) = \mathbf{P}(Q_{i_1 j_1} \in S) \cdots \mathbf{P}(Q_{i_n j_n} \in S) \lesssim_{d,n} N_{m+1}^{-dn}. \quad (15)$$

Taking expectations over all cubes in  $B$ , and applying (15) gives

$$\begin{aligned} \mathbf{E}(\#(\mathcal{Q}_{m+1}(B) \cap \mathcal{Q}_{m+1}(S^n))) &\lesssim_{d,n} \#(\mathcal{Q}_{m+1}(B)) \cdot N_{m+1}^{-dn} \\ &\leq l_{m+1}^{-(s+\varepsilon)} N_{m+1}^{-dn} \\ &= \frac{M_{m+1}^{s+\varepsilon} l_m^{-(s+\varepsilon)}}{N_{m+1}^{dn-s-\varepsilon}}. \end{aligned} \quad (16)$$

If (8) holds, for an appropriately chosen constant depending only on  $l_m, d, n$ , and  $s$ , we can apply Markov's inequality together with (9) and (16) to conclude

$$\begin{aligned} \mathbf{P}(\mathcal{Q}_{k+1}(B) \cap \mathcal{Q}_{k+1}(S^n) \neq \emptyset) &= \mathbf{P}(\#(\mathcal{Q}_{k+1}(B) \cap \mathcal{Q}_{k+1}(S^n)) \geq 1) \\ &\leq \mathbf{E}(\#(\mathcal{Q}_{k+1}(B) \cap \mathcal{Q}_{k+1}(S^n))) \\ &\leq 1/N_{m+1}^\varepsilon \leq 1/10. \end{aligned} \quad (17)$$

Thus  $\mathcal{Q}_{k+1}(S^n)$  is disjoint from  $\mathcal{Q}_{k+1}(B)$  with high probability.

Now we analyze the Fourier transform of the measure  $\nu$ . For each  $i \in \Pi_{m+1}^d$ , and  $m \in \mathbf{Z}$ , define  $X_{im} = \widehat{\nu}_i(m) - \widehat{\mathbf{E}(\nu_i)}(m)$ . Note that

$$\begin{aligned} \sum_{i \in \Pi_{m+1}^d} \widehat{\mathbf{E}(\nu_i)}(m) &= \sum_{i \in \Pi_{m+1}^d} l_{m+1}^d \sum_{j \in [N_{m+1}]^d} \int_{\mathbf{R}^d} e^{-2\pi i m \cdot x} \psi_{m+1}(x - a_{ij}) d\mu_T(x) \\ &= \int_{\mathbf{R}^d} e^{-2\pi i m \cdot x} d\mu_T(x) = \widehat{\mu_T}(m). \end{aligned} \quad (18)$$

For each  $i$  and  $m$ , the standard  $(L^1, L^\infty)$  bound on the Fourier transform, combined with (11), shows

$$\begin{aligned} \|X_{im}\|_{\psi_2(L)} &\leq \|X_{im}\|_{L^\infty} \leq \|\nu_i(\mathbf{R}^d)\|_{L^\infty} + \mathbf{E}(\nu_i)(\mathbf{R}^d) \\ &\lesssim_d \mathbf{E}(\nu_i)(\mathbf{R}^d) + r_{m+1}^{d+1} \|\nabla \mu_T\|_{L^\infty(\mathbf{R}^d)}. \end{aligned} \quad (19)$$

For a fixed  $m$ , the family of random variables  $\{X_{im}\}$  are independent. Furthermore,  $\sum X_{im} = \widehat{\nu}(m) - \widehat{\mathbf{E}(\nu)}(m)$ , and

$$\begin{aligned} \mathbf{E}(\widehat{\nu}_S(m)) &= \frac{r_{m+1}^d}{N_{m+1}^d} \int_{\mathbf{R}^d} e^{-2\pi i m \cdot x} \left( \sum_{(i,j) \in \Sigma_{m+1}^d} \psi_{m+1}(x - a_{ij}) \right) d\mu_T(x) \\ &= \frac{r_{m+1}^d l_{m+1}^{-d}}{N_{m+1}^d} \int_{\mathbf{R}^d} e^{-2\pi i m \cdot x} d\mu_T(x) \\ &= \frac{r_{m+1}^d l_{m+1}^{-d}}{N_{m+1}^d} \widehat{\mu_T}(m) = \widehat{\mu_T}(m). \end{aligned}$$

Thus we may apply Hoeffding's inequality to (19) to conclude that

$$\|\widehat{\nu}(m) - \widehat{\mu_T}(m)\|_{\psi_2(L)} \lesssim_d \left( \sum \mathbf{E}(\nu_i)(\mathbf{R}^d)^2 \right)^{1/2} + r_{m+1}^{d/2+1} \|\nabla \mu_T\|_{L^\infty(\mathbf{R}^d)}.$$

Now taking in absolute values gives the inequality

$$\begin{aligned} \mathbf{E}(\nu_i(\mathbf{R}^d)) &= l_{m+1}^d \sum_{j \in [N_{m+1}]^d} \int \psi_{m+1}(x - a_{ij}) d\mu_T(x) \\ &\leq r_{m+1}^d \|\mu_T\|_{L^\infty(\mathbf{R}^d)}, \end{aligned}$$

and so

$$\|\widehat{\nu}(m) - \widehat{\mu_T}(m)\|_{\psi_2(L)} \lesssim_d [\|\mu_T\|_{L^\infty(\mathbf{R}^d)} + \|\nabla \mu_T\|_{L^\infty(\mathbf{R}^d)}] r_{m+1}^{d/2}. \quad (20)$$

If (8) holds, for an appropriately chosen constant depending only on  $l_m, d, n$ , and  $s$ , then we can apply a union bound over  $D = \{m \in \mathbf{Z}^d : |m| \leq 10l_{m+1}^{-1}\}$  to conclude that there exists a constant  $c(\mu_T, d)$  such that

$$\begin{aligned} \mathbf{P}\left(\|\widehat{\nu} - \widehat{\mu_T}\|_{L^\infty(D)} \geq r_{m+1}^{d/2} \log(M_{m+1})\right) &\lesssim_d l_{m+1}^{-d} \exp\left(-c(\mu_T, d) \log(M_{m+1})^2\right) \\ &\leq 1/10. \end{aligned}$$

Thus  $\widehat{\nu}$  and  $\widehat{\mu_T}$  are highly likely to differ only by a miniscule amount over small frequencies.

Finally, it suffices to analyze the values of  $\widehat{\nu}_S(m)$  when  $|m| \geq 10l_{m+1}^{-1}$ . We note that if we define a random measure

$$\alpha = r_{m+1}^d \sum_{\substack{i \in \Pi_{m+1}^d \\ d(a_i, T) \leq 2r_{m+1}^{-1}}} \delta_{a_{ij_i}},$$

then  $\nu_S = (\alpha * \psi_{m+1})\mu_T$ . Thus we have  $\widehat{\nu}_S = \widehat{\mu_T} * (\widehat{\alpha} \cdot \widehat{\psi_{m+1}})$ . Since  $\mu_T$  is compactly supported, we can define, for each  $t > 0$ ,

$$A(t) = \sup |\widehat{\mu_T}(\xi)| |\xi|^t < \infty.$$

Similarly, since

$$\widehat{\psi_{m+1}}(\xi) = \widehat{\psi}(l_{m+1}^{-1}\xi),$$

if we set, for each  $t > 0$ ,

$$B(t) = \sup |\widehat{\psi}(\xi)| |\xi|^t,$$

then

$$\sup |\widehat{\psi_{m+1}}(\xi)| |\xi|^t = l_{m+1}^{-t} B(t).$$

It now suffices to bound

$$\sup_{|\eta| \geq 10l_{m+1}^{-1}} |\eta|^{s/2} \int \widehat{\mu_T}(\eta - \xi) \widehat{\alpha}(\xi) \widehat{\psi_{m+1}}(\xi) d\xi.$$

Since  $\alpha(\mathbf{R}^d) \leq 2^d$ ,  $\|\widehat{\alpha}\|_{L^\infty(\mathbf{R}^d)} \leq 2^d$ , so it suffices to understand

$$\int |\widehat{\mu_T}(\eta - \xi)| |\widehat{\psi_{m+1}}(\xi)| d\xi.$$

If  $|\xi| \leq |\eta|/2$ ,  $|\eta - \xi| \geq |\xi|/2$ , so for all  $t > 0$ ,

$$\int_{0 \leq |\xi| \leq |\eta|/2} |\widehat{\mu_T}(\eta - \xi)| |\widehat{\psi_{m+1}}(\xi)| d\xi \leq \frac{A(t) 2^{t-d}}{|\eta|^{t-d}}.$$

If we set  $t = d + 1 + s/2$  and apply (8) for an appropriate chosen constant depending only on  $d$  and  $\mu_T$ , we conclude

$$\int_{0 \leq |\xi| \leq |\eta|/2} |\widehat{\mu_T}(\eta - \xi)| |\widehat{\psi_{m+1}}(\xi)| d\xi \leq \frac{A(d + 1 + s/2) 2^{1+s/2} |\eta|^{-1}}{|\eta|^{s/2}} \leq \frac{|\eta|^{-s/2}}{10 \cdot 2^d}.$$

Conversely, if  $|\xi| \geq 2|\eta|$ , then  $|\eta - \xi| \geq |\xi|/2$ , so for each  $t > 0$ ,

$$\begin{aligned} \int_{|\xi| \geq 2|\eta|} |\widehat{\mu_T}(\eta - \xi)| |\widehat{\psi_{m+1}}(\xi)| d\xi &\leq \int_{|\xi| \geq 2|\eta|} \frac{A(t) 2^t}{|\xi|^t} \\ &\lesssim_d \int_{2|\eta|}^\infty r^{d-t} A(t) l_{m+1}^{-t'} 2^t. \end{aligned}$$

Provided  $t > d + 1$ , this integral is finite, and is

$$\lesssim_{d,t} A(t) 2^{d+1} |\eta|^{d+1-t}.$$

Setting  $t = d + 2 + s/2$ , and applying (8), we conclude

$$\int_{|\xi| \geq 2|\eta|} |\widehat{\mu_T}(\eta - \xi)| |\widehat{\psi_{m+1}}(\xi)| d\xi \leq \frac{|\eta|^{-s/2}}{10 \cdot 2^d}.$$

Finally, we conclude that for each  $t > 0$ ,

$$\int_{|\eta|/2 \leq |\xi| \leq 2|\eta|} |\widehat{\mu_T}(\eta - \xi)| |\widehat{\psi_{m+1}}(\xi)| d\xi \leq \frac{2^d |\eta|^d B(t) 2^t}{|\xi|^t}.$$

If we set  $t = d + s/2 + 1$  and apply (8), we conclude

$$\int_{|\eta|/2 \leq |\xi| \leq 2|\eta|} |\widehat{\mu_T}(\eta - \xi)| |\widehat{\psi_{m+1}}(\xi)| d\xi \leq \frac{|\xi|^{-s/2}}{10 \cdot 2^d}.$$



Summing up the three bounds, we conclude that if  $|\eta| \geq 10l_{m+1}^{-1}$ , then

$$|\widehat{\nu_S}(\eta)| \leq |\eta|^{-s/2}.$$

□