*Proof of Theorem II.2.* The proof follows ideas from [18, Theorem 3.20] or [1, Theorem 3.2] and utilizes the discrete subsampling from [2]. By [2, Lemma 2.2 and Theorem 3.1] we obtain

$$\frac{n}{2} \le \sigma_{\min}^2(\mathbf{L}) = \|(\mathbf{L}^* \mathbf{L})^{-1} \mathbf{L}^*\|_{2 \to 2}^2$$
(5)

with probability  $1 - \exp(-t)$ . We split the approximation error as follows

$$||f - S_{\mathbf{X}} \mathbf{f}||_{L_{2}}^{2} = ||f - P_{I} f||_{L_{2}}^{2} + ||P_{I} f - S_{\mathbf{X}} \mathbf{f}||_{L_{2}}^{2}$$

$$\leq ||f - P_{I} f||_{L_{2}}^{2} + 2||P_{I} f - S_{\mathbf{X}} P_{I_{M}} \mathbf{f}||_{L_{2}}^{2} + 2||S_{\mathbf{X}} P_{I_{M}} f - S_{\mathbf{X}} \mathbf{f}||_{L_{2}}^{2}.$$

By event (5) and the invariance of  $S_X$  to functions supported on I, we obtain

$$||f - S_{\mathbf{X}}f||_{L_{2}}^{2} \leq ||f - P_{I}f||_{L_{2}}^{2} + \frac{4}{n} \sum_{i \in J} |(P_{I}f - P_{I_{M}}f)(\mathbf{x}^{i})|^{2} + \frac{4}{n} \sum_{i \in J} |(f - P_{I_{M}}f)(\mathbf{x}^{i})|^{2}$$

$$\leq ||f - P_{I}f||_{L_{2}}^{2} + \frac{4}{n} \sum_{i \in J} |(P_{I}f - P_{I_{M}}f)(\mathbf{x}^{i})|^{2} + 4||f - P_{I_{M}}f||_{\infty}^{2}.$$

$$(6)$$

We will estimate the middle summand by Bernstein's inequality, cf. [9, Corollary 7.31] or [24, Theorem 6.12]. We define random variables  $\xi_i = |(P_I f - P_{I_M} f)(\boldsymbol{x}^i)|^2 - \|P_{I_M} f - P_I f\|_{L_2}^2$ . By the reconstructing property, we have  $\frac{1}{M} \sum_{j=1}^M |(P_I f - P_{I_M} f)(\boldsymbol{x}^i)|^2 = \|P_{I_M} f - P_I f\|_{L_2}^2$  and, thus,  $\xi_i$  are mean zero. Further, we have

$$\|\xi_i\|_{\infty} \le (\|P_I f - P_{I_M} f\|_{\infty} + \|P_I f - P_{I_M} f\|_{L_2})^2$$

and

$$\mathbb{E}(\xi_i^2) = \mathbb{E}\Big(\Big(|(P_I f - P_{I_M} f)(\boldsymbol{x}^i)|^2\Big)^2\Big) + \|P_I f - P_{I_M} f\|_{L_2}^4$$

$$\leq \|P_I f - P_{I_M} f\|_{\infty}^2 \sum_{i=1}^M |(P_I f - P_{I_M} f)(\boldsymbol{x}^i)|^2 + \|P_I f - P_{I_M} f\|_{L_2}^4$$

$$\leq \|P_I f - P_{I_M} f\|_{L_2}^2 (\|P_I f - P_{I_M} f\|_{\infty} + \|P_I f - P_{I_M} f\|_{L_2})^2.$$

Applying Bernstein's inequality yields

$$\frac{1}{n} \sum_{i \in J} |(P_I f - P_{I_M} f)(\boldsymbol{x}^i)|^2 - \|P_I f - P_{I_M} f\|_{L_2}^2 \le \frac{2t}{3n} (\|P_I f - P_{I_M} f\|_{\infty} + \|P_I f - P_{I_M} f\|_{L_2})^2 \\
+ \sqrt{\frac{2t}{n}} \|P_I f - P_{I_M} f\|_{L_2} (\|P_I f - P_{I_M} f\|_{\infty} + \|P_I f - P_{I_M} f\|_{L_2})$$

with probability  $1 - \exp(-t)$ .

Plugging this in (6) and using  $||P_{I_M}f - P_If||_{L_2} \le ||f - P_If||_{L_2}$ , we obtain

$$||f - S_{\mathbf{X}} \mathbf{f}||_{L_{2}}^{2} \leq 5||f - P_{I} f||_{L_{2}}^{2} + 2\sqrt{\frac{8t}{n}}||f - P_{I} f||_{L_{2}} \left(||P_{I_{M}} f - P_{I} f||_{\infty} + ||f - P_{I} f||_{L_{2}}\right)$$

$$+ \frac{8t}{3n} \left(||P_{I_{M}} f - P_{I} f||_{\infty} + ||f - P_{I} f||_{L_{2}}\right)^{2} + 4||f - P_{I_{M}} f||_{\infty}^{2}$$

$$\leq \left(\sqrt{5}||f - P_{I} f||_{L_{2}} + \sqrt{\frac{8t}{3n}} \left(||P_{I_{M}} f - P_{I} f||_{\infty} + ||f - P_{I} f||_{L_{2}}\right)\right)^{2}$$

$$+ 4||f - P_{I_{M}} f||_{\infty}^{2}.$$

Using the assumption on n, we have  $8t/3n \le 2/9|I| \le 2/9$  and further estimate

$$||f - S_{\mathbf{X}} \mathbf{f}||_{L_2}^2 \le \left(3||f - P_I f||_{L_2} + \sqrt{\frac{2}{9|I|}} ||P_{I_M} f - P_I f||_{\infty}\right)^2 + 4||f - P_{I_M} f||_{\infty}^2.$$

The second part of the bound is obtained by applying Hölder's inequality:

$$||P_{I_M}f - P_If||_{\infty}^2 = \left\| \sum_{\mathbf{k} \in I_M \setminus I} \hat{f}_{\mathbf{k}} \exp(2\pi i \langle \mathbf{k}, \cdot \rangle) \right\|_{\infty}^2$$

$$\leq \left\| \sum_{\mathbf{k} \in I_M \setminus I} |\exp(2\pi i \langle \mathbf{k}, \cdot \rangle)|^2 \right\|_{\infty} \sum_{\mathbf{k} \in I_M \setminus I} |\hat{f}_{\mathbf{k}}|^2$$

$$= |I_M \setminus I| ||P_{I_M}f - P_If||_{L_2}^2 \leq |I_M \setminus I| ||f - P_If||_{L_2}^2.$$

By union bound we obtain the overall probability.

**Lemma A.1.** For X uniformly distributed on  $\mathbb{T}^d$ , we have

$$\mathbb{P}(|g(X)| \ge \delta') \ge \frac{\|g\|_{L_2}^2 - \delta'^2}{\|g\|_{\infty}^2}.$$

*Proof.* Consequence of [19, Par. 9.3.A] for  $h(t) = t^2$ .

Proof of Theorem III.1. From Lemma A.1 follows

$$\mathbb{P}(|\hat{f}_{\{1,...,t\},k}(\boldsymbol{\xi})| \geq \delta') \geq \frac{\|\hat{f}_{\{1,...,t\},k}\|_{L_2}^2 - \delta'^2}{\|\hat{f}_{\{1,...,t\},k}\|_{\infty}^2} =: p_k.$$

Let  $\overline{p} = \max_{k \in \mathcal{P}_{\{1,...,t\}}(I_{\delta})} p_k$ . By union bound, we have the successful detection with probability exceeding  $1 - |I_{\delta}|(1 - \overline{p})^r$ . This is smaller or equal  $1 - \varepsilon$  whenever

$$r \le -\frac{\log|I_{\delta}| + \log(1/\varepsilon)}{\log(1-\overline{p})}. \tag{7}$$

Since  $p_k \in (0,1)$ , we obtain

$$-\frac{1}{\log(1-p_{k})} \leq \frac{1}{p_{k}} = \frac{\|\hat{f}_{\{1,\dots,t\},k}\|_{\infty}^{2}}{\|\hat{f}_{\{1,\dots,t\},k}\|_{L_{2}}^{2} - \delta'^{2}} \leq \frac{\left(\sum_{(k,l) \in \mathbb{Z}^{d-t}} |\hat{f}_{(k,l)}|\right)^{2}}{\sum_{(k,l) \in \mathbb{Z}^{d-t}} |\hat{f}_{(k,l)}|^{2} - \delta'^{2}}.$$

With  $\mathbf{k} \in \mathcal{P}_{\{1,\dots,t\}}(I_{\delta})$  we have

$$\delta'^2 \le \delta^2/2 \le \frac{1}{2} \sum_{(k,l) \in \mathbb{Z}^{d-t}} |\hat{f}_{(k,l)}|^2.$$

Consequently,

$$-\frac{1}{\log(1-p_{k})} \leq 2 \frac{\left(\sum_{(k,l)\in\mathbb{Z}^{d-t}} |\hat{f}_{(k,l)}|\right)^{2}}{\sum_{(k,l)\in\mathbb{Z}^{d-t}} |\hat{f}_{(k,l)}|^{2}}$$

$$\leq 4 \frac{\left(\sum_{(k,l)\in\mathcal{P}_{\{1,...,t\},k}(I_{\delta})} |\hat{f}_{(k,l)}|\right)^{2}}{\sum_{(k,l)\in\mathbb{Z}^{d-t}} |\hat{f}_{(k,l)}|^{2}} + 4 \frac{\left(\sum_{(k,l)\notin\mathcal{P}_{\{1,...,t\},k}(I_{\delta})} |\hat{f}_{(k,l)}|\right)^{2}}{\sum_{(k,l)\in\mathbb{Z}^{d-t}} |\hat{f}_{(k,l)}|^{2}}$$

$$\leq 4 |\mathcal{P}_{\{1,...,t\},k}(I_{\delta})| + \frac{4}{\delta^{2}} \left(\sum_{(k,l)\notin\mathcal{P}_{\{1,...,t\},k}(I_{\delta})} |\hat{f}_{(k,l)}|\right)^{2}$$

$$\leq 4 |I_{\delta}| + \frac{4}{\delta^{2}} \left(\sum_{l\notin I_{\delta}} |\hat{f}_{l}|\right)^{2}.$$

Plugging this into (7) we obtain the assertion.

Proof of Theorem III.2. By Theorem III.1 we have

$$\begin{split} \frac{\delta}{\sqrt{2}} &\leq \max_{i=1,\dots,r} |\hat{f}_{\{1,\dots,t\},\boldsymbol{k}}(\boldsymbol{\xi}^i)| \\ &\leq \max_{i=1,\dots,r} |\hat{f}_{\{1,\dots,t\},\boldsymbol{k}}(\boldsymbol{\xi}^i) - \hat{g}_{\{1,\dots,t\},\boldsymbol{k}}(\boldsymbol{\xi}^i)| + |\hat{g}_{\{1,\dots,t\},\boldsymbol{k}}(\boldsymbol{\xi}^i)| \,. \end{split}$$

Next, we show a bound on the above difference in order to obtain the desired threshold. We have

$$\begin{split} |\hat{f}_{\{1,...,t\},\boldsymbol{k}}(\boldsymbol{\xi}^{i}) - \hat{g}_{\{1,...,t\},\boldsymbol{k}}(\boldsymbol{\xi}^{i})|^{2} &\leq \sum_{k \in I_{\{1,...,t\}}} |\hat{f}_{\{1,...,t\},\boldsymbol{k}}(\boldsymbol{\xi}^{i}) - \hat{g}_{\{1,...,t\},\boldsymbol{k}}(\boldsymbol{\xi}^{i})|^{2} \\ &\leq \|f(\cdot,\boldsymbol{\xi}^{i}) - S_{\boldsymbol{X}}f(\cdot,\boldsymbol{\xi}^{i}\|_{L_{2}}^{2} \,. \end{split}$$

Using Theorem 2.2 and the assumption on the carnality of the frequency index sets, we obtain

$$\begin{split} &|\hat{f}_{\{1,\dots,t\},\boldsymbol{k}}(\boldsymbol{\xi}^{i}) - \hat{g}_{\{1,\dots,t\},\boldsymbol{k}}(\boldsymbol{\xi}^{i})|^{2} \\ &\leq \Big(3 + \sqrt{\frac{2|I_{\{1,\dots,t\}}^{M} \setminus I_{\{1,\dots,t\}}|}{9|I_{\{1,\dots,t\}}|}}\Big)^{2} \|f(\cdot,\boldsymbol{\xi}^{i}) - P_{I_{\{1,\dots,t\}}}f(\cdot,\boldsymbol{\xi}^{i})\|_{L_{2}}^{2} \\ &\quad + 4\|f(\cdot,\boldsymbol{\xi}^{i}) - P_{I_{\{1,\dots,t\}}^{M}}f(\cdot,\boldsymbol{\xi}^{i})\|_{\infty}^{2} \\ &\leq 16\|f(\cdot,\boldsymbol{\xi}^{i}) - P_{I_{\{1,\dots,t\}}}f(\cdot,\boldsymbol{\xi}^{i})\|_{L_{2}}^{2} + 4\|f(\cdot,\boldsymbol{\xi}^{i}) - P_{I_{\{1,\dots,t\}}^{M}}f(\cdot,\boldsymbol{\xi}^{i})\|_{\infty}^{2} \end{split}$$

with probability

$$1 - 2r \exp(-t) \quad \text{and} \quad |\pmb{X}| \geq 12 |I_{\{1,\dots,t\}}| (\log |I_{\{1,\dots,t\}}| + t) \,,$$

where union bound was used over the detection iterations. Setting  $t = \log(2r/\varepsilon)$  we obtain the assertion.