Density of sampling and interpolation in reproducing kernel Hilbert spaces

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Abstract
We derive necessary density conditions for sampling and for interpolation in general reproducing kernel Hilbert spaces satisfying some natural conditions on the geometry of the space and the reproducing kernel. If the volume of shells is small compared to the volume of balls (weak annular decay property) and if the kernel possesses some off-diagonal decay or even some weaker form of localization, then there exists a critical density $D$ with the following property: a set of sampling has density $\geq D$, whereas a set of interpolation has density $\leq D$. The main theorem unifies many known density theorems in signal processing, complex analysis, and harmonic analysis. For the special case of bandlimited function we recover Landau’s fundamental density result. In complex analysis we rederive a critical density for generalized Fock spaces. In harmonic analysis we obtain the first general result about the density of coherent frames.

1. Introduction
How many samples of a function $f$ are necessary to completely recover $f$? The first answer is the sampling theorem of Whittaker, Kotelnikov, Shannon, and others [51]. It provides an explicit and elegant reconstruction formula for the recovery of a bandlimited function from its samples on a grid and establishes a fundamental relation between the bandwidth of $f$ and the sampling density (the Nyquist rate in engineering terminology). This sampling theorem is the basis of modern information theory [51] and remains the model for analog-digital and digital-analog conversion.

The decisive mathematical theorems work for more general notions of bandwidth and for non-uniform sampling and are due to Beurling [7, 46] (sufficient conditions) and Landau [32]. Landau’s necessary conditions give a precise meaning to the concept of a Nyquist rate for bandlimited functions.

To this day, Landau’s theorem is the prototype of a density theorem, it has inspired several hundred papers on sampling. Landau’s necessary conditions have been transferred, modified, and adapted to dozens of similar situations. Here is a short, but by no means exhaustive list of density theorems in the wake of Landau.

(i) Sampling in spaces of analytic functions, in particular, in Bargmann–Fock space [34, 44, 47] and in generalized Fock spaces [1, 33, 38].
(ii) Sampling of bandlimited functions with derivatives [25, 35].
(iii) Necessary density conditions of Gabor frames [13, 42]. This topic alone has attracted about hundred papers, for a detailed history of this density theorem we refer to Heil’s survey article [26].

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(iv) Density conditions for abstract frames with some localization properties [4–6].
(v) Sampling in spaces of bandlimited functions on Lie groups [20, 39].
(vi) Sampling in spaces of variable bandwidth [24].
(vii) Sampling in spaces of bandlimited functions associated to an integral transform, for example, the Hankel transform [2].
(viii) Density of frames in the orbit of an irreducible unitary representation of a homogeneous nilpotent Lie group [27].

Essentially each of these contributions on necessary density conditions for sampling and interpolation uses and modifies one of these three methods.

(i) Landau’s original method is based on the spectral analysis of a family of localization operators (composition of the projection onto bandlimited functions with a time-limiting operator). This method is very powerful, but can become quite technical. Usually, the generalization of Landau’s method is difficult.

(ii) The method of Ramanathan and Steger [42] was originally developed to prove the density theorem for Gabor frames. It compares and estimates the dimension of finite-dimensional subspaces corresponding to a local patch of a sampling set with the dimension of finite-dimensional subspaces corresponding to a local patch of an interpolating set. This is the method used most frequently. However, it is not universally applicable, because it requires the existence of a set that is simultaneously sampling and interpolating (or at least the construction of interpolating sets and sampling sets with almost the same density).

(iii) A third method goes back to Kolountzakis and Lagarias [30] (proof of Lemma 2.3) who studied the density of tilings by translation. This method was then used by Iosevich and Kolountzakis [28] to prove a version of Landau’s theorem and by Nitzan–Olevski [37] to give the simplest proof of Landau’s theorem. With hindsight this method consists of comparing a set of sampling to a continuous frame. For us this is the method of choice that we will use in our investigation.

We observe that all density theorems listed above treat certain Hilbert spaces with a reproducing kernel (or isomorphic copies thereof). This fact and the similarity of all proofs raise the question of a universal density theorem in reproducing kernel Hilbert spaces (RKHS). This point of view leads immediately to the pertinent questions: what are the concrete conditions on the underlying configuration space and on the reproducing kernel to lead to a density theorem? What is the relevant density concept in a reproducing kernel Hilbert space? Is there a critical density in a reproducing kernel Hilbert space that separates sets of sampling from sets of interpolation?

In this paper we attempt to give an answer to these questions and will prove a general density theorem for functions in a reproducing kernel Hilbert space. Here is a simplified version of our main result.

**Theorem 1.1.** Let $X$ be a metric measure space with a metric $d$ and a measure $\mu$ such that balls have finite measure, $\mu$ is non-degenerate, and satisfies the weak annular decay property, that is, $\inf_x \mu(B_r(x)) > 0$ for some $r > 0$, and

$$\lim_{r \to \infty} \sup_{x \in X} \frac{\mu(B_r(x) \setminus B_{r-1}(x))}{\mu(B_r(x))} = 0.$$ 

Furthermore, let $\mathcal{H} \subseteq L^2(X, \mu)$ be a reproducing kernel Hilbert space with a reproducing kernel $k(x, y)$ satisfying $\inf_{x \in X} k(x, x) > 0$ and an off-diagonal decay condition of the form

$$|k(x, y)| \leq C (1 + d(x, y))^{-\sigma} \text{ for all } x, y \in X$$

for some $\sigma > 0$ satisfying $\lim_{r \to \infty} \sup_{x \in X} \int_{X \setminus B_r(x)} (1 + d(x, y))^{-2\sigma} d\mu(y) = 0.$

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[7] **Theorem 1.1.** Let $X$ be a metric measure space with a metric $d$ and a measure $\mu$ such that balls have finite measure, $\mu$ is non-degenerate, and satisfies the weak annular decay property, that is, $\inf_x \mu(B_r(x)) > 0$ for some $r > 0$, and

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for some $\sigma > 0$ satisfying $\lim_{r \to \infty} \sup_{x \in X} \int_{X \setminus B_r(x)} (1 + d(x, y))^{-2\sigma} d\mu(y) = 0.$
(i) Necessary conditions for sampling: If for \( \Lambda \subset X \) there exist \( A, B > 0 \) such that
\[
A \|f\|^2 \leq \sum_{\lambda \in \Lambda} |f(\lambda)|^2 \leq B \|f\|^2 \quad \text{for all } f \in \mathcal{H},
\]
then
\[
D^-(\Lambda) := \liminf_{r \to \infty} \inf_{x \in X} \frac{\#(\Lambda \cap B_r(x))}{\mu(B_r(x))} \geq \liminf_{r \to \infty} \inf_{x \in X} \frac{1}{\mu(B_r(x))} \int_{B_r(x)} k(y, y) d\mu(y).
\]

(ii) Necessary conditions for interpolation: Likewise, let \( \Lambda \subset X \) and assume that for every \( a \in \ell^2(\Lambda) \), there exists a function \( f \in \mathcal{H} \) such that
\[
f(\lambda) = a_\lambda, \quad \lambda \in \Lambda,
\]
then
\[
D^+(\Lambda) := \limsup_{r \to \infty} \sup_{x \in X} \frac{\#(\Lambda \cap B_r(x))}{\mu(B_r(x))} \leq \limsup_{r \to \infty} \sup_{x \in X} \frac{1}{\mu(B_r(x))} \int_{B_r(x)} k(y, y) d\mu(y).
\]

Following established terminology, we call a set \( \Lambda \subset X \) that satisfies the sampling inequality (2) a set of (stable) sampling, while a set satisfying the interpolation property in part (ii) of Theorem 1.1 is called a set of interpolation. Alternatively, \( \Lambda \) is a set of sampling, if and only if \( \{k_\lambda : \lambda \in \Lambda\} \) is a frame for \( \mathcal{H} \), and \( \Lambda \) is a set of interpolation if and only if \( \{k_\lambda : \lambda \in \Lambda\} \) is a Riesz sequence in \( \mathcal{H} \), that is, if there exist \( A, B > 0 \) such that
\[
A \|c\|^2 \leq \left\| \sum_{\lambda \in \Lambda} c_\lambda k_\lambda \right\|^2 \leq B \|c\|^2 \quad \text{for all } c \in \ell^2(\Lambda).
\]

The densities \( D^-(\Lambda) \) and \( D^+(\Lambda) \) are the obvious generalizations of the lower and upper Beurling density to metric spaces.

The principal merit of Theorem 1.1 is the clarification of the main notions that go into a density theorem. To prove a density theorem, one needs

(i) geometric data and the compatibility of metric and measure, and

(ii) estimates for the reproducing kernel.

The verification of these properties is by no means trivial. Indeed, kernel estimates (Bergman, Bargmann, and other reproducing kernels) constitute a deep and rich area of analysis. Theorem 1.1 shifts the emphasis in proofs of density theorems: it is important to understand the geometry and the reproducing kernel, but it is no longer necessary to prove a ‘new’ density theorem from scratch with tedious modifications of known techniques.

As an example we show how Landau’s original theorem follows from Theorem 1.1. The discussion also shows some of the difficulties in applying Theorem 1.1.

Let \( X = \mathbb{R}^d \) with Lebesgue measure \( \mu \) and \( \Omega \subset \mathbb{R}^d \) be a set of finite Lebesgue measure and define \( B_\Omega = \{f \in L^2(\mathbb{R}^d) : \text{supp} \hat{f} \subset \Omega\} \) to be the subspace of band-limited functions with spectrum in \( \Omega \). Then
\[
f(x) = \int_\Omega \hat{f}(\xi) e^{2\pi i x \cdot \xi} d\xi = \int_{\mathbb{R}^d} f(y) \int_\Omega e^{2\pi i \xi \cdot (x - y)} d\xi dy,
\]
and therefore \( B_\Omega \) is a reproducing kernel Hilbert space with reproducing kernel \( k(x, y) = \int_\Omega e^{2\pi i \xi \cdot (x - y)} d\xi = \hat{1}_\Omega(y - x) \). Clearly \( \mathbb{R}^d \) satisfies all geometric conditions of Theorem 1.1. The computation of the averaged trace is equally easy, since \( k(x, x) = \int_\Omega 1 d\xi = \mu(\Omega) \) and thus
\[
\frac{1}{\mu(B_r(x))} \int_{B_r(x)} k(y, y) d\mu(y) = \mu(\Omega) \quad \text{independent of } x \text{ and } r.
\]
Thus a set of sampling must satisfy \( D^-(\Lambda) \geq \mu(\Omega) \) and a set of interpolation \( D^+(\Lambda) \leq \mu(\Omega) \), which is Landau’s theorem. The decay condition (1) is satisfied for simple spectra, for example, when \( \Omega \) is a cube or a convex set with
smooth boundary. Yet we must be cautious: in general the kernel does not satisfy the decay condition in (1), because the Fourier transform of the characteristic function of a compact set may decay arbitrarily slowly \[21\]. In the main theorem (Theorem 2.2) we will impose a much weaker condition on the kernel and thus will take care of this subtle point.

Most density theorems of the list above can be understood as an example of the general density theorem in reproducing kernel Hilbert spaces. To demonstrate the wide applicability of Theorem 1.1 we will rederive some of the fundamental density theorems in several areas of analysis.

(i) **Signal analysis:** As already indicated, Theorem 1.1 implies Landau’s necessary density conditions for bandlimited functions.

(ii) **Complex analysis** (in several variables): We will deduce Lindholm’s density conditions \[33\] for generalized Fock spaces.

(iii) **Harmonic analysis:** We will derive a necessary condition for the density of a frame in the orbit of a square-integrable, unitary representation of a group of polynomial growth. A special case of this result is the density theorem for Gabor frames.

Theorem 1.1 is definitely not the end of density theorems. Of course, it is our main ambition to prove new density results. In this sense the axiomatic set-up serves as a preparation for future work. Currently there are numerous results about sampling theorems for ‘sufficiently dense’ sets. See, for instance, \[15, 29, 39–41\]. These results need to be complemented by a critical density, provided that it exists at all.

Finally, let us point out some limitations. We mention that the weak annular decay property of the measure is not always satisfied, as it is tied to the growth of balls in \(X\). Thus Theorem 1.1 excludes a number of very interesting examples, for instance, density theorems in Bergman spaces \[43, 45, 46\] and the density of wavelet frames \[31\]. However, in these cases the Beurling density is not the correct density, and to this date it is an open problem whether a critical density always exists in this context. In our view the Beurling density is the correct notion of density in geometries with polynomial (or subexponential) growth, whereas in geometries with exponential growth new phenomena arise (which are not at all understood). See Section 5.6 for a hint.

The paper is organized as follows: In Section 2 we collect the set of assumptions for a general density theorem and then formulate the main result. Section 3 contains a discussion of the main hypotheses and several technical lemmas. In Section 4 we provide the proof of the main density theorem. In Section 5 we apply the density theorem to rederive several fundamental density theorems from the literature.

### 2. The density theorem

#### 2.1. Assumptions

We first list the general assumptions on the geometry and the reproducing kernel to make a density theorem work.

(A) **Assumptions on the metric and the measure.** We assume that \((X, d)\) is a metric space, and \(\mu\) is a measure on \(X\) with the following properties:

- The metric \(d : X \times X \to [0, +\infty)\) is a \(\mu \otimes \mu\) measurable function and the balls \(B_r(x) = \{y \in X : d(y, x) < r\}\) satisfy \(\mu(B_r(x)) < \infty\) for all \(r > 0\) and \(x \in X\).

- **Non-degeneracy of balls** (Axiom NDB): There exists \(r > 0\) such that

  \[
  \inf_{x \in X} \mu(B_r(x)) > 0. \tag{4}
  \]
Weak annular decay property (Axiom WAD): Spherical shells have small volume compared to full balls:
\[
\lim_{r \to \infty} \sup_{x \in X} \frac{\mu(B_{r+1}(x) \setminus B_r(x))}{\mu(B_r(x))} = 0.
\] (5)

(B) Assumptions on the reproducing kernel. We assume that \(H \subseteq L^2(X, \mu)\) is a reproducing kernel Hilbert space with reproducing kernel \(k(x, y)\) so that
\[
f(x) = \int_X k(x, y)f(y)d\mu(y) = \langle f, k_x \rangle,
\]
where \(k_x(y) = k(y, x) = k(x, y)\). The assumptions on the kernel are as follows:

• Condition on the diagonal (Axiom D): there exist constants \(C_1, C_2 > 0\) such that for all \(x \in X\)
\[
C_1 \leq k(x, x) \leq C_2.
\] (6)

• Weak localization of the kernel (Axiom WL): For every \(\epsilon > 0\) there is a constant \(r = r(\epsilon)\), such that
\[
\sup_{x \in X} \int_{X \setminus B_r(x)} |k(x, y)|^2d\mu(y) < \epsilon^2.
\] (7)

• The homogeneous approximation property (Axiom HAP): Assume that \(\Lambda\) is such that \(\{k_\lambda : \lambda \in \Lambda\}\) is a Bessel sequence for \(H\), that is, \(\sum_{\lambda \in \Lambda} |f(\lambda)|^2 \leq C\|f\|^2\) for all \(f \in H\). Then for every \(\epsilon > 0\) there is a constant \(r = r(\epsilon)\), such that
\[
\sup_{x \in X} \sum_{\lambda \in \Lambda \setminus B_r(x)} |k(x, \lambda)|^2 < \epsilon^2.
\] (8)

We note that this version of axiom (HAP) differs from the usual homogeneous approximation property used in the literature. Compare the discussion in [4].

2.2. Sets, densities and traces

A set \(\Lambda \subseteq X\) is called relatively separated if there exists \(\rho_0 > 0\) such that for all \(\rho \geq \rho_0\), there exists \(C = C_\rho > 0\) such that:
\[
\#(\Lambda \cap B_\rho(x)) \leq C\mu(B_\rho(x)), \quad x \in X.
\] (9)

**Definition 2.1.** The lower Beurling density of a set \(\Lambda \subseteq X\) is defined to be
\[
D^-(\Lambda) = \liminf_{r \to \infty} \inf_{x \in X} \frac{\#(\Lambda \cap B_r(x))}{\mu(B_r(x))}.
\] (10)

and the upper Beurling density of \(\Lambda\) is
\[
D^+(\Lambda) = \limsup_{r \to \infty} \sup_{x \in X} \frac{\#(\Lambda \cap B_r(x))}{\mu(B_r(x))}.
\] (11)

Note that for relatively separated sequences, the upper Beurling density is finite by the non-degeneracy of the balls (4).
We will compare the density of a set of sampling or interpolation to an invariant of the reproducing kernel Hilbert space. The correct Nyquist rate is the averaged trace of the kernel. We define the lower and upper traces as follows:

\[ \text{tr}^-(k) = \liminf_{r \to \infty} \inf_{x \in X} \frac{1}{\mu(B_r(x))} \int_{B_r(x)} k(y,y) d\mu(y), \]

\[ \text{tr}^+(k) = \limsup_{r \to \infty} \sup_{x \in X} \frac{1}{\mu(B_r(x))} \int_{B_r(x)} k(y,y) d\mu(y). \]

2.3. Main result

In this general context one can prove the following necessary density conditions for sets of sampling and interpolation.

**Theorem 2.2.** Assume that \( \mathcal{H} \) is a reproducing kernel Hilbert space of functions on a metric measure space \( X \) satisfying the assumptions of Section 2.1.

(i) If \( \Lambda \subseteq X \) is a set of stable sampling for \( \mathcal{H} \), then

\[ D^-(\Lambda) \geq \text{tr}^-(k) \quad \text{and} \quad D^+(\Lambda) \geq \text{tr}^+(k). \]

(ii) If \( \Lambda \subseteq X \) is a set of interpolation for \( \mathcal{H} \), then

\[ D^-(\Lambda) \leq \text{tr}^-(k) \quad \text{and} \quad D^+(\Lambda) \leq \text{tr}^+(k). \]

It is instructive to work out the trivial case of a space \( X \) with finite diameter, that is, \( X = B_R(x) \) for all \( x \in X \). Assuming the existence of a set \( \Lambda \) of sampling in this case, we observe by Lemma 3.7 below that \( \Lambda \) is finite, hence \( \mathcal{H} \) is finite-dimensional. Choosing any orthonormal basis \( \varphi_1, \ldots, \varphi_m \) of \( \mathcal{H} \), we have

\[ k(x,y) = \sum_{j=1}^m \varphi_j(x) \overline{\varphi_j(y)}. \]

In particular, we get

\[ \text{tr}^-(k) = \text{tr}^+(k) = \frac{1}{\mu(X)} \int_X k(y,y) d\mu(y) = \frac{m}{\mu(X)}. \]

On the other hand, given any subset \( \Lambda \), we have

\[ D^+(\Lambda) = D^-(\Lambda) = \frac{\# \Lambda}{\mu(X)}. \]

Hence Theorem 2.2 says that \( \# \Lambda \leq m \) is a necessary condition for sets of interpolation, whereas \( \# \Lambda \geq m \) is necessary for sets of sampling (=uniqueness). Of course, both results follow from an elementary dimension count: The cardinality of a Riesz sequence is bounded by the dimension of the underlying vector space, whereas a frame must contain a basis and its cardinality exceeds the dimension of \( \mathcal{H} \).

3. Discussion of the assumptions and preliminary lemmas

In the following we always assume the axioms from Section 2.1. We discuss the axioms and prove some easy consequences.
3.1. Metric and measure

Most of the geometric conditions are technical conditions to exclude pathologies.

First note that we do not assume that the $\sigma$-algebra of $\mu$-measurable sets—the domain of $\mu$—is generated by the class of open sets associated with the metric $d$. Rather, we only assume that the function $d$ is $\mu \otimes \mu$ measurable. This means that the sets $\{(x, y) \in X \times X : d(x, y) < r\}$ belong to the smallest $\sigma$-algebra generated by the sets $A \times B$, with $A, B$ in the domain of $\mu$. In particular, the balls $B_r(x)$ are $\mu$-measurable. However, more general sets that are open with respect to the metric $d$ may not be $\mu$-measurable, since they may fail to be a countable union of balls.

The decisive condition is the weak annular decay property (WAD). This condition links the metric and the measure and imposes some compatibility between them. Even in simple examples, axiom (WAD) requires some care. For example, the standard metric $d_1(x, y) = |x - y|$ on $\mathbb{R}$ with Lebesgue measure fulfills the weak annular decay property, but the topologically equivalent metric $d_2(x, y) = \log(1 + |x - y|)$ violates (WAD).

**Remark 3.1.** For complicated geometries the verification of the weak annular decay property is decidedly non-trivial. In the literature one often uses the stronger annular decay property [48]. A metric measure space $(X, d, \mu)$ satisfies this property, if there exist constants $C > 0$ and $\delta \in (0, 1]$ such that for every $h \in [0, 1]$, $r > 0$, $x \in X$,

$$
\mu(B_r(x) \setminus B_{r(1-h)}(x)) \leq C h^\delta \mu(B_r(x)).
$$

The annular decay property implies the weak annular decay property. Indeed, if (16) holds, the choice $h = r^{-1}$ shows that $\mu(B_r(x) \setminus B_{r-1}(x)) \leq C r^{-\delta} \mu(B_r(x))$. In addition, for $r \geq 1$ and $\varepsilon > 0$,

$$
\mu(B_r(x)) \leq \mu(B_{r+\varepsilon}(x)) \leq C r^\delta (r + \varepsilon)^{-\delta} \mu(B_{r}(x))
$$

So, choosing $\varepsilon$ small enough ($\varepsilon = (2C)^{-1/\delta}$), we see that $\mu(B_r(x)) \leq C \mu(B_r(x))$. Similarly, we see that $\mu(B_r(x)) \leq C' \mu(B_{r-1}(x))$, for $r \gg 1$ and a constant $C' > 0$, so the conclusion follows.

In the literature one finds several conditions that imply the annular decay property, for instance, if $X$ is a length space or if $X$ has monotone geodesics. See [9, 48] for further discussion and more information on the annular decay property. Axiom (WAD) seems to be tied to the growth of balls and seems compatible with at most subexponential growth.

We now note that the measure $\mu$ is locally doubling at large scales.

**Lemma 3.2.** There exists $r_0 > 0$ such that for all $r \geq r_0$, there is a constant $C_r > 0$ such that

$$
\mu(B_{2r}(x)) \leq C_r \mu(B_r(x)) \quad \text{for all } x \in X.
$$

**Proof.** By Axiom (WAD), there exists $r_0$, such that for $r \geq r_0$ and all $x \in X$, $\mu(B_{r+1}(x) \setminus B_r(x)) \leq \mu(B_r(x))$. As a consequence, $\mu(B_{r+1}(x)) \leq 2 \mu(B_r(x))$. Iterating this estimate we conclude that $\mu(B_{2r}(x)) \leq C_r \mu(B_r(x))$, where $C_r := 2^{[r]}$.

**Remark 3.3.** Note that the proof of Lemma 3.2 only depends on Axiom (WAD).

The locally doubling property in (17) is much weaker than the usual doubling property for measures. See [12, 49] for more on locally doubling spaces.
We next formulate two lemmas on the number of points in and the measure of general ‘spherical’ shells.

**Lemma 3.4.** Let $\Lambda \subseteq X$ be relatively separated. Then for all sufficiently large $\rho > 0$, and $R > \rho$, $r > 0$, $x \in X$, we have

$$\#(\Lambda \cap (B_{R+r}(x) \setminus B_R(x))) \leq C_{\rho,\Lambda} \mu(B_{R+r+\rho}(x) \setminus B_{R-\rho}(x)),$$

where the constant $C_{\rho,\Lambda}$ depends only on $\rho$ and $\Lambda$.

**Proof.** By Lemma 3.2, for sufficiently large $\rho$, the locally doubling property

$$\mu(B_{\rho}(x)) \leq C_{\rho/2} \mu(B_{\rho/2}(x))$$

(19)

and the estimate in (9) hold.

Let $\{B_{\rho/2}(y) : y \in X_0\}$ be a maximal packing of $B_{R+r}(x) \setminus B_R(x)$. This means that (i) $X_0 \subseteq B_{R+r}(x) \setminus B_R(x)$, (ii) $\{B_{\rho/2}(y) : y \in X_0\}$ is a disjoint family of balls and (iii) the family is maximal with respect to the properties (i) and (ii). By maximality, $\{B_{\rho}(y) : y \in X_0\}$ is a covering of $B_{R+r}(x) \setminus B_R(x)$. Using (9) and (19), we obtain

$$\#(\Lambda \cap (B_{R+r}(x) \setminus B_R(x))) \leq \# \left( \bigcup_{y \in X_0} \Lambda \cap B_{\rho}(y) \right) \leq C \sum_{y \in X_0} \mu(B_{\rho}(y))$$

$$\leq CC_{\rho/2} \sum_{y \in X_0} \mu(B_{\rho/2}(y)) = CC_{\rho/2} \mu \left( \bigcup_{y \in X_0} B_{\rho/2}(y) \right)$$

$$\leq CC_{\rho/2} \mu(B_{R+r+\rho}(x) \setminus B_{R-\rho}(x)),$$

where $C$ provided by (9) depends only on $\rho$ and $\Lambda$. \hfill $\square$

We will apply the weak annular decay property in the following versions.

**Lemma 3.5.** If $(X, d, \mu)$ satisfies the weak annular decay property, then, for all $\rho' > 0$,

$$\lim_{r \to \infty} \sup_{x \in X} \frac{\mu(B_{r+\rho'}(x))}{\mu(B_r(x))} = 1,$$

(20)

and

$$\lim_{r \to \infty} \sup_{x \in X} \frac{\mu(B_{r+\rho'}(x) \setminus B_{r-\rho'}(x))}{\mu(B_r(x))} = 0.$$  

(21)

**Proof.** For the proof of (20), we observe that weak annular decay is equivalent to

$$\lim_{r \to \infty} \sup_{x \in X} \frac{\mu(B_{r+1}(x))}{\mu(B_r(x))} = 1.$$  

$$\lim_{r \to \infty} \sup_{x \in X} \frac{\mu(B_{r+1}(x))}{\mu(B_r(x))} = 1.$$  

For given $\rho' > 0$, we set $N = \lceil \rho' \rceil$ and obtain

$$1 \leq \sup_{x \in X} \frac{\mu(B_{r+\rho'}(x))}{\mu(B_r(x))} \leq \sup_{x \in X} \frac{\mu(B_{r+N}(x))}{\mu(B_r(x))}$$

$$\leq \prod_{j=1}^{N} \sup_{x \in X} \frac{\mu(B_{r+j}(x))}{\mu(B_{r+j-1}(x))} \rightarrow 1, \text{ as } r \to \infty.$$

(21)
This proves (20). For the proof of (21) note that
\[
0 \leq \sup_{x \in X} \frac{\mu(B_{r+\rho}(x) \setminus B_{r-\rho}(x))}{\mu(B_r(x))} \\
\leq \sup_{x \in X} \frac{\mu(B_{r+\rho}(x))}{\mu(B_r(x))} - \inf_{x \in X} \frac{\mu(B_{r-\rho}(x))}{\mu(B_r(x))}
\]
which converges to \(1 - 1 = 0\) by (20), as \(r \to \infty\). \(\square\)

3.2. The reproducing kernel

In practice, the upper bound in Axiom D (6) and the weak localization property follow from off-diagonal decay estimates for the kernel (see Section 5). To verify the lower bound in Axiom (D), the following observation can be useful.

**Lemma 3.6.** Let \(k\) be a reproducing kernel on \(X \times X\). If there is a constant \(C > 0\) such that for all \(x \in X\), there exists \(f_x \in X\) such that \(\|f_x\| \leq C\) and \(f_x(x) = 1\), then the lower bound in Axiom (D) holds.

**Proof.** The claim follows from \(1 = f_x(x) = \langle f_x, k_x \rangle \leq C\|k_x\| = Ck(x, x)^{1/2}\). \(\square\)

**Normalization of the reproducing kernel.** In some examples, for example, in Fock spaces of entire functions, the reproducing kernel is unbounded. This situation can be dealt with by the following normalization. Let \(\psi : X \to \mathbb{R}^+\) and define a new measure \(\tilde{\mu} = \psi^2 \mu\). Then \(J\) defined by \(Jf = f\psi^{-1}\) is an isometry from \(L^2(X, \mu)\) onto \(L^2(X, \tilde{\mu})\) and \(\mathcal{H} = \mathcal{H} \subseteq L^2(X, \tilde{\mu})\) is again a reproducing kernel Hilbert space. We calculate the new reproducing kernel \(\tilde{k}\) as follows:

\[
Jf(x) = \psi(x)^{-1}f(x) = \psi(x)^{-1}\langle f, k_x \rangle = \psi(x)^{-1}\langle Jf, Jk_x \rangle,
\]
whence the new reproducing kernel is

\[
\tilde{k}(x, y) = \psi(x)^{-1}\psi(y)^{-1}k(x, y) \quad x, y \in X.
\] (22)

In particular, if we choose \(\psi(x) = \|k_x\|\), then \(\tilde{k}(x, y) = \|k_x\|^{-1}\|k_y\|^{-1}k(x, y)\) and thus \(\tilde{k}(x, x) = 1\). Axiom D (6) can therefore always be fulfilled by a renormalization of the kernel. However, this may be at the price of destroying some other required properties. Note that in this normalization the critical density is always one, provided that Theorem 2.2 is still applicable. See also [10] for normalization of weighted Bergman spaces on the disk.

**Lemma 3.7.** Let \(\Lambda \subseteq X\) be a set such that \(\{k_\lambda : \lambda \in \Lambda\}\) is a Bessel sequence in \(\mathcal{H}\). Then \(\Lambda\) is relatively separated.

**Proof.** Let \(\epsilon^2 := \frac{1}{2} \inf_{x \in X} k(x, x)\). By Axiom D (6), \(\epsilon > 0\). Select \(r_1 = r_1(\epsilon)\) according to Axiom (WL). Since \(\int_X |k(x, y)|^2 d\mu(y) = \|k_x\|^2 = k(x, x)\), we obtain that, for \(\rho \geq r_1\),

\[
\int_{B_\rho(x)} |k(x, y)|^2 d\mu(y) = k(x, x) - \int_{X \setminus B_\rho(x)} |k(x, y)|^2 d\mu(y) \geq \epsilon^2.
\] (23)

By hypothesis, there exists a constant \(C > 0\) such that for all \(f \in \mathcal{H}\)

\[
\sum_{\lambda \in \Lambda} |f(\lambda)|^2 \leq C\|f\|^2,
\]
and this holds in particular for $f = k_y$. Using Lemma 3.2, select $r_0$ such that (17) holds for $r \geq r_0$. Let $x \in X$ and $\rho \geq \rho_0 := \max\{r_0, r_1\}$. Using (23) and Axiom D (6), we estimate
\[
\epsilon^2 \#(\Lambda \cap B_{\rho}(x)) \leq \sum_{\lambda \in \Lambda \cap B_{\rho}(x)} |k(\lambda, y)|^2 d\mu(y)
\leq \sum_{\lambda \in \Lambda \cap B_{\rho}(x)} \int_{B_{2\rho}(x)} |k(\lambda, y)|^2 d\mu(y) \leq \int_{B_{2\rho}(x)} \sum_{\lambda \in \Lambda} |k_y(\lambda)|^2 d\mu(y)
\leq C \sup_{y \in X} \|k_y\|^2 \mu(B_{2\rho}(x)) \leq CC\rho \mu(B_{\rho}(x)).
\]
Hence, (9) holds. $\square$

**Remark 3.8.** Note that the proof of Lemma 3.7 does not use Axiom (HAP).

We will need the following elementary facts about frames and Riesz sequences in a reproducing kernel Hilbert space $\mathcal{H}$. They are copied from [24, 37].

**Lemma 3.9.** The following properties hold.

(i) Assume that $\{k_\lambda : \lambda \in \Lambda\}$ is a frame for $\mathcal{H}$ with canonical dual frame $\{g_\lambda : \lambda \in \Lambda\}$. Then $k_\lambda$ and $g_\lambda$ satisfy the following:
\[
\sum_{\lambda \in \Lambda} k_\lambda(y)g_\lambda(y) = k(y, y), \quad y \in X,
\]  
(24)
\[
\sup_{y \in X} \sum_{\lambda \in \Lambda} |g_\lambda(y)|^2 < \infty, \quad y \in X,
\]  
(25)
\[
\sup_{\lambda \in \Lambda} \|g_\lambda\| = C < \infty,
\]  
(26)
\[
\sup_{\lambda \in \Lambda} |\langle k_\lambda, g_\lambda \rangle| \leq 1.
\]  
(27)

(ii) If $\{k_\lambda : \lambda \in \Lambda\}$ is a Riesz basis for a subspace $V \subseteq \mathcal{H}$ with biorthogonal basis $\{g_\lambda : \lambda \in \Lambda\} \subseteq V$, then (25), (26) hold true, while (24) is replaced by the inequality
\[
0 \leq \sum_{\lambda \in \Lambda} k_\lambda(y)g_\lambda(y) \leq k(y, y),
\]  
(28)
and in (27) holds the equality
\[
\langle g_\lambda, k_\lambda \rangle = 1 \quad \text{for all } \lambda \in \Lambda.
\]  
(29)

**Proof.** The proof from [24, 37] is included for completeness. Let $P_V$ be the orthogonal projection on the subspace $V$ of $\mathcal{H}$. Inequality (28) follows from
\[
\sum_{\lambda \in \Lambda} k_\lambda(y)g_\lambda(y) = \sum_{\lambda \in \Lambda} \langle k_\lambda, k_y \rangle \langle k_y, g_\lambda \rangle
\leq \sum_{\lambda \in \Lambda} \langle k_y, g_\lambda \rangle k_\lambda, k_y \rangle
= \langle P_V k_y, k_y \rangle
\leq \|k_y\|^2 = k(y, y).
The proof of (24) is the same, except that $P_V = I$ for frames and thus equality holds in the last step.

Item (25) follows from

$$\sum_{\lambda \in \Lambda} |g_\lambda(y)|^2 = \sum_{\lambda \in \Lambda} |(g_\lambda, k_y)|^2 \leq C\|k_y\|^2 = Ck(y, y),$$

where $C$ is the upper frame bound for $\{g_\lambda : \lambda \in \Lambda\}$, and $k(y, y)$ is uniformly bounded by Axiom D (6).

Item (27) is an immediate consequence of the minimality of the $\ell^2$-norms of the coefficients in the canonical frame expansion [14]:

$$k_{\lambda'} = \sum_{\lambda \in \Lambda} (k_{\lambda'}, g_\lambda)k_\lambda = 1 \cdot k_{\lambda'} \quad \text{for every } \lambda' \in \Lambda,$$

so

$$|\langle k_{\lambda'}, g_{\lambda'}\rangle|^2 \leq \sum_{\lambda \in \Lambda} |\langle k_{\lambda'}, g_\lambda\rangle|^2 \leq 1 \quad \text{for every } \lambda' \in \Lambda.$$

Finally (26) is a general fact about frames.

\[\square\]

4. Proof of Theorem 2.2

In this section we prove the necessary density conditions for sets of sampling or interpolation in general reproducing kernel Hilbert spaces satisfying the conditions of Section 2.1. Similar to [28, 37], our proof is inspired by the method of Kolountzakis and Lagarias [30]. It is modeled on our own version in [24].

**Proof.** Step 1. The first part of the proof works both for sets of sampling and for sets of interpolation. Equivalently, we assume that $\{k_\lambda : \lambda \in \Lambda\}$ is either a frame with canonical dual frame $\{g_\lambda : \lambda \in \Lambda\}$ or that $\{k_\lambda : \lambda \in \Lambda\}$ is a Riesz sequence for some subspace $V \subseteq H$ with biorthogonal basis $\{g_\lambda : \lambda \in \Lambda\} \subseteq V$. By Lemma 3.7, the set $\Lambda$ is relatively separated. Let $\rho > 0$ be a suitably large radius, such that the conclusion of Lemma 3.4 holds.

In both cases we estimate the quantity

$$\int_{B_r(x)} \sum_{\lambda \in \Lambda} k_\lambda(y)\overline{g_\lambda(y)}d\mu(y) \quad (30)$$

for large $r$.

Fix $\epsilon > 0$ and $x \in X$, and choose $R = R(\epsilon)$ such that both kernel axioms WL (7) and HAP (8) are satisfied. In the proof, we will just write $B_r$ for the ball $B_r(x)$ to abbreviate the notation.

We partition $\Lambda$ and write accordingly

$$\sum_{\lambda \in \Lambda} k_\lambda(y)g_\lambda(y) = \left( \sum_{\lambda \in \Lambda \cap B_{r-R}} + \sum_{\lambda \in \Lambda \cap (X \setminus B_{r+R})} + \sum_{\lambda \in \Lambda \cap (B_{r+R} \setminus B_{r-R})} \right) k_\lambda(y)g_\lambda(y)$$

$$= A_1(y) + A_2(y) + A_3(y).$$

**Estimate of $\int_{B_r} A_1$.** We estimate $|\int_{B_r} A_1(y)d\mu(y)|$. We write

$$\int_{B_r} A_1(y)d\mu(y) = \int_X \sum_{\lambda \in \Lambda \cap B_{r-R}} k_\lambda(y)g_\lambda(y)d\mu(y) - \int_{X \setminus B_r} \sum_{\lambda \in \Lambda \cap B_{r-R}} k_\lambda(y)g_\lambda(y)d\mu(y),$$
and set
\[ L = \sum_{\lambda \in \Lambda \cap B_{r-R}} \int_{X \setminus B_r} k_\lambda(y)g_\lambda(y) d\mu(y). \]

Then
\[ \int_{B_r} A_1(y) d\mu(y) = \sum_{\lambda \in \Lambda \cap B_{r-R}} \langle k_\lambda, g_\lambda \rangle - L. \]  

If \( \lambda \in \Lambda \cap B_{r-R} \) and \( y \in X \setminus B_r \), then \( d(\lambda, y) > R \). Therefore the kernel axiom WL (7) and (26) imply that a single term contributing to \( L \) is majorized by
\[ \left| \int_{X \setminus B_r} k_\lambda(y)\overline{g_\lambda(y)} d\mu(y) \right| \leq \left( \int_{X \setminus B_R(\lambda)} |k_\lambda(y)|^2 d\mu(y) \right)^{1/2} \|g_\lambda\| \leq \epsilon' C'' \].

This estimate implies
\[ |L| \leq \epsilon C_1 \#(\Lambda \cap B_{r-R}) \leq \epsilon C_1 \#(\Lambda \cap B_r). \]  

**Estimate of \( \int_{B_r} A_2 \).** Note that \( y \in B_r \) and \( \lambda \in \Lambda \setminus B_{r+R} \) implies that \( d(\lambda, y) > R \). Then Axiom HAP (8) ensures that \( \sum_{\lambda \in \Lambda \setminus B_{r+R}} |k(y, \lambda)|^2 \leq \sum_{\lambda \in \Lambda \setminus B_{r-R}} |k(y, \lambda)|^2 < C^2 \). Consequently, using also (25), we obtain
\[ \left| \int_{B_r} A_2(y) d\mu(y) \right| \leq \int_{B_r} \left( \sum_{\lambda \in \Lambda \setminus B_{r+R}} |k_\lambda(y)|^2 \right)^{1/2} \left( \sum_{\lambda \in \Lambda} |g_\lambda(y)|^2 \right)^{1/2} d\mu(y) \leq C_2 \mu(B_r). \]  

**Estimate of \( \int_{B_r} A_3 \).** For the third term observe that
\[ \int_{B_r} |A_3(y)| d\mu(y) \leq \sum_{\lambda \in \Lambda \setminus (B_{r+R} \setminus B_{r-R})} \int_X |k_\lambda(y)| |g_\lambda(y)| d\mu(y) \leq \sum_{\lambda \in \Lambda \setminus (B_{r+R} \setminus B_{r-R})} \|k_\lambda\| \|g_\lambda\|. \]

Using Axiom D (6), and the boundedness of the canonical dual frame (26), we obtain
\[ \int_{B_r} |A_3(y)| d\mu(y) \leq C_3 \#(\Lambda \cap (B_{r+R} \setminus B_{r-R})). \]  

From now on we distinguish the case of sets of sampling from sets of interpolation.

**Step 2.** Assume first that \( \{k_\lambda : \lambda \in \Lambda\} \) is a Riesz sequence in \( \mathcal{H} \). We rewrite the expansion
\[ \int_{B_r} \sum_{\lambda \in \Lambda} k_\lambda(y)\overline{g_\lambda(y)} d\mu(y) = \int_{B_r} \sum_{j=1}^3 A_j(y) d\mu(y), \]  

and, with the help of (31), (28), and (29), we obtain the estimate
\[ \#(\Lambda \cap B_{r-R}) = \sum_{\lambda \in \Lambda \cap B_{r-R}} \langle k_\lambda, g_\lambda \rangle \]
\[ = \int_{B_r} A_1(y) d\mu(y) + L \]
Lemma 3.4 bounds the last term by

\[ - (1 + \epsilon) C \mu(B_{r+R} \setminus B_{r-R}) \]

so we conclude that

\[ (1 - \epsilon C_1) \frac{\#(\Lambda \cap B_r)}{\mu(B_r)} \leq \frac{1}{\mu(B_r)} \int_{B_r} k(y, y) \, d\mu(y) + \epsilon C_2 + (1 + C_3) C_{\rho, \Lambda} \frac{\mu(B_{r+R} \setminus B_{r-R})}{\mu(B_r)}. \]

We recall that \( B_r = B_r(x) \), take the supremum over all \( x \in X \), let \( r \) tend to \( \infty \), and use Lemma 3.5 to deduce

\[ (1 - \epsilon C_1) D^+(\Lambda) \leq \text{tr}^+(k) + \epsilon C_2 + (1 + C_3) C_{\rho, \Lambda} \limsup_{r \to \infty} \sup_{x \in X} \frac{\mu(B_{r+R} \setminus B_{r-R} \setminus x)}{\mu(B_r(x))} \]

\[ = \text{tr}^+(k) + \epsilon C_2. \]

Since \( \epsilon > 0 \) is arbitrary, it follows that \( D^+(\Lambda) \leq \text{tr}^+(k) \) for every interpolating set \( \Lambda \). The inequality \( D^-(\Lambda) \leq \text{tr}^-(k) \) follows from (38) in a similar way, just taking inf instead of sup and lim sup instead of lim sup.

**Step 3.** Assume next that \( \{k_\lambda : \lambda \in \Lambda\} \) is a frame for \( \mathcal{H} \). Then by Lemma 3.9, (24) and (27), we have

\[ \sum_{\lambda \in \Lambda} k_\lambda(y) g_\lambda(y) = k(y, y) \]

and

\[ |\langle k_\lambda, g_\lambda \rangle| \leq 1. \]

Proceeding as in Step 2 we obtain with \( x \in X \) fixed and \( B_r = B_r(x) \) that

\[ \int_{B_r} k(y, y) \, d\mu(y) = \int_{B_r} \sum_{\lambda \in \Lambda} k_\lambda(y) g_\lambda(y) \, d\mu(y) = \sum_{j=1}^{3} A_j(y) \, d\mu(y) \]

\[ = \sum_{\lambda \in \Lambda \cap B_{r-R}} \langle k_\lambda, g_\lambda \rangle - L + \int_{B_r} A_2(y) \, d\mu(y) + \int_{B_r} A_3(y) \, d\mu(y) \]

\[ \leq \#(\Lambda \cap B_{r-R}) + \epsilon C_1 \#(\Lambda \cap B_r) + \epsilon C_2 \mu(B_r) + C_3 \#(\Lambda \cap (B_{r+R} \setminus B_{r-R})) \]

\[ \leq (1 + \epsilon C_1) \#(\Lambda \cap B_r) + \epsilon C_2 \mu(B_r) + C_3 C_{\rho, \Lambda} \mu(B_{r+R} \setminus B_{r-R}). \]
Consequently
\[
\frac{1}{\mu(B_r)} \int_{B_r} k(y, y) d\mu(y) \leq (1 + \epsilon C_1) \frac{\#(\Lambda \cap B_r)}{\mu(B_r)} + \epsilon C_2 + C_3 C_{\rho, \Lambda} \frac{\mu(B_{r+R+\rho} \setminus B_{r-R-\rho})}{\mu(B_r)}.
\] (39)

Again, we take the infimum over all \( x \in X \) and let \( r \) tend to \( \infty \) to obtain via Lemmas 3.4 and 3.5
\[
\text{tr}^-(k) \leq (1 + \epsilon C_1) D^-(\Lambda) + \epsilon C_2.
\]
Since \( \epsilon > 0 \) was arbitrary, the necessary density is \( D^-(\Lambda) \geq \text{tr}^-(k) \), as claimed. As in Step 2, the statement involving the upper trace and density follows by just taking sup instead of inf and lim sup instead of lim inf. \( \square \)

By drawing a different conclusion at the end of the above proof, the density theorem can be given a dimension-free form as suggested to us by J. Ortega-Cerdà.

**Corollary 4.1.** Impose the same assumption on \((X, d, \mu)\) and \(\mathcal{H}\) as in Theorem 2.2.

(i) If \( \Lambda \subseteq X \) is a set of stable sampling for \(\mathcal{H}\), then
\[
\liminf_{r \to \infty} \inf_{x \in X} \frac{\#(\Lambda \cap B_r(x))}{\int_{B_r(x)} k(y, y) d\mu(y)} \geq 1.
\] (40)

(ii) If \( \Lambda \subseteq X \) is a set of interpolation for \(\mathcal{H}\), then
\[
\limsup_{r \to \infty} \sup_{x \in X} \frac{\#(\Lambda \cap B_r(x))}{\int_{B_r(x)} k(y, y) d\mu(y)} \leq 1.
\] (41)

**Proof.** We only prove (i), as (ii) is similar. Dividing (39) yields
\[
1 \leq (1 + \epsilon C_1) \frac{\#(\Lambda \cap B_r)}{\int_{B_r} k(y, y) d\mu(y)} + \epsilon C_2 \frac{\mu(B_r)}{\int_{B_r} k(y, y) d\mu(y)} + C_3 C_{\rho, \Lambda} \frac{\mu(B_{r+R+\rho} \setminus B_{r-R-\rho})}{\int_{B_r} k(y, y) d\mu(y)}.
\]
Since \( \int_{B_r} k(y, y) d\mu(y) \geq C_1 \mu(B_r) \) by (6), the second term on the right-hand side is of order \( \epsilon \), and the third term tends to 0 for \( r \to \infty \). Taking the infimum over all \( x \in X \) and letting \( r \) tend to \( \infty \), we obtain
\[
1 \leq (1 + C_1 \epsilon) \liminf_{r \to \infty} \inf_{x \in X} \frac{\#(\Lambda \cap B_r(x))}{\int_{B_r(x)} k(y, y) d\mu(y)} + \epsilon C',
\]
which yields assertion (i). \( \square \)

This corollary suggests that one could define the modified Beurling density of a set \( \Lambda \) by (40) and (41). The corresponding density theorem is then dimension-free with critical density 1 independent of the geometry of the reproducing kernel Hilbert space. By contrast, the critical density in Theorem 2.2 depends on the reproducing kernel.

**Remark 4.2.** Instead of the Beurling densities one may also apply an ultra-filter to (38) and (39) and obtain a density theorem with respect to a so-called frame measure function. See [5, 6] for the notion of frame measure function and its applications to the comparison of frames.

**Remark 4.3.** Theorem 2.2(ii) is valid without axiom (HAP).

Our proof of the density theorem emphasized the symmetry between sampling and interpolation. We have seen that the same estimates are used in both density theorems. If we give
Lemma 3.2—which, as noted in Remark 3.3, only depends on Axiom (WAD)—to further grant we can invoke that lemma to obtain \( \rho > 0 \) such that (9) holds. Similarly, we can invoke Lemma 3.7—on which readily yields

\[
\#(\Lambda \cap B_r) = \sum_{\lambda \in \Lambda \cap B_r} \langle k_{\lambda}, Pg_{\lambda} \rangle = \left( \int_{B_{r+R}} + \int_{X \setminus B_{r+R}} \right) \left( \sum_{\lambda \in \Lambda \cap B_r} k_{\lambda}(y)Pg_{\lambda}(y) \right) d\mu(y) = I + \tilde{L}.
\]

By (32) and (33) we obtain \( |\tilde{L}| \leq \epsilon C_1 \#(\Lambda \cap B_r) \), whereas (28) yields

\[
I = \int_{B_{r+R}} k(y, y) d\mu(y) \leq \int_{B_r} k(y, y) d\mu(y) + \sup_{x \in X} k(y, y) \mu(B_{r+R} \setminus B_r).
\]

Consequently,

\[
\#(\Lambda \cap B_r) \leq \int_{B_r} k(y, y) d\mu(y) + C_0 \mu(B_{r+R} \setminus B_r) + \epsilon C_1 \#(\Lambda \cap B_r),
\]

which readily yields

\[
D^\pm(\Lambda) \lesssim \text{tr}^\pm(k).
\]

4.1. Off-diagonal decay with respect to a metric

In applications, the reproducing kernel often possesses some off-diagonal decay. In this case the kernel axioms are easier to check. The following proposition shows that Theorem 1.1 is a special case of Theorem 2.2.

Proof of Theorem 1.1. We show that the hypotheses of Section 2.1 are satisfied. The assumption on \( d \) and \( k \) clearly implies the weak localization condition (WL) and the diagonal condition (D). It only remains to check the homogeneous approximation property (HAP). Assume that \( \Lambda \subseteq X \) is such that \( \{k_{\lambda} : \lambda \in \Lambda\} \) is a Bessel sequence.

As noted in Remark 3.8, the proof of Lemma 3.7 does not depend on Axiom (HAP). Hence we can invoke that lemma to obtain \( \rho > 0 \) such that (9) holds. Similarly, we can invoke Lemma 3.2—which, as noted in Remark 3.3, only depends on Axiom (WAD)—to further grant that

\[
\mu(B_{2\rho}(x)) \leq C_{\rho} \mu(B_{\rho}(x)), \quad x \in X.
\]

We observe first that the obvious inequality

\[
1 + d(x, y) \leq (1 + d(x, \lambda))(1 + d(\lambda, y)) \quad \text{for all} \quad x, y, \lambda \in X
\]

implies that

\[
(1 + d(x, \lambda))^{-2\sigma} \leq (1 + d(\lambda, y))^{2\sigma}(1 + d(x, y))^{-2\sigma}.
\]

Therefore,

\[
|k(x, \lambda)|^2 \leq C(1 + d(x, \lambda))^{-2\sigma} = \frac{C}{\mu(B_{\rho}(\lambda))} \int_{B_{\rho}(\lambda)} (1 + d(x, \lambda))^{-2\sigma} d\mu(y)
\]

\[
\leq \frac{C}{\mu(B_{\rho}(\lambda))} \int_{B_{\rho}(\lambda)} (1 + d(\lambda, y))^{2\sigma}(1 + d(x, y))^{-2\sigma} d\mu(y)
\]

\[
\leq C \frac{(1 + \rho)^{2\sigma}}{\mu(B_{\rho}(\lambda))} \int_{B_{\rho}(\lambda)} (1 + d(x, y))^{-2\sigma} d\mu(y).
\]
Consequently,
\[
\sum_{\lambda \in \Lambda \cap (X \setminus B_r(x))} |k(x, \lambda)|^2 \\
\leq C(1 + \rho)^{2\sigma} \int_X \left( \sum_{\lambda \in \Lambda \cap B_r(x)} \mu(B_\rho(\lambda))^{-1} \mathbf{1}_{B_\rho(\lambda)}(y) \right) (1 + d(x, y))^{-2\sigma} \, d\mu(y).
\]

We note that the sum vanishes if \( d(x, y) \leq r - \rho \), thus the integral can be taken over the set \( X \setminus B_{r-\rho}(x) \). Next we estimate the sum for fixed \( y \in X \setminus B_{r-\rho}(x) \). Note that if \( y \in B_\rho(\lambda) \), then \( B_\rho(y) \subseteq B_{2\rho}(\lambda) \), and, by (42),
\[
\mu(B_\rho(\lambda)) \geq C_\rho^{-1} \mu(B_{2\rho}(\lambda)) \geq C_\rho^{-1} \mu(B_\rho(y)).
\]

Hence, using (9), we can estimate
\[
\sum_{\lambda \in \Lambda \cap B_r(x)} \frac{1}{\mu(B_\rho(\lambda))} \mathbf{1}_{B_\rho(\lambda)}(y) \leq \frac{C_\rho}{\mu(B_\rho(y))} \sum_{\lambda \in \Lambda} \mathbf{1}_{B_\rho(y)}(\lambda) \leq CC_\rho.
\]

In conclusion,
\[
\sum_{\lambda \in \Lambda \cap B_r(x)} |k(x, \lambda)|^2 \leq CC_\rho (1 + \rho)^{2\sigma} \int_{X \setminus B_{r-\rho}(x)} (1 + d(x, y))^{-2\sigma} \, d\mu(y).
\]

By hypothesis, this expression tends to zero uniformly in \( x \) as \( r \to \infty \), whence \( k \) satisfies (HAP). \( \square \)

5. Examples

In this section we discuss several examples of density theorems from different areas of analysis. Our point is to show that some of the fundamental density theorems in signal analysis, complex analysis, frame theory, and harmonic analysis follow from the axiomatic approach. All we have to do is to check the general conditions of Section 2.1 and formulate the corresponding theorem. This is not always easy, and our discussion will point out some of the difficulties and pitfalls.

5.1. Bandlimited functions

Let \( \Omega \subseteq \mathbb{R}^d \) be measurable with finite Lebesgue measure and \( B_\Omega = \{ f \in L^2(\mathbb{R}^d) : \text{supp} \, \hat{f} \subseteq \Omega \} \) be the corresponding Paley–Wiener space. As observed in the introduction, its reproducing kernel is
\[
k(x, y) = \int_{\Omega} e^{-2\pi i \xi \cdot (y-x)} \, d\xi.
\]

Clearly \( X = \mathbb{R}^d \) with the Euclidean distance and Lebesgue measure \( d\mu(x) = dx \) satisfy the geometrical assumptions. As for the kernel, we have \( k(x, x) = |\Omega| \) and thus the averaged trace is obviously \( \text{tr}^+(k) = \text{tr}^-(k) = \mu(\Omega) \).

The verification of the weak localization of the kernel is easy, because \( k(x, y) = \int_{\Omega} (y-x) \, d\xi \) where \( \int_{\Omega} \) is the Fourier transform of an \( L^2 \)-function. Therefore
\[
\int_{\mathbb{R}^d \setminus B_r(x)} \left| \int_{\Omega} (y-x) \, d\xi \right|^2 \, dy = \int_{|y| \geq r} \left| \int_{\Omega} (y-x) \, d\xi \right|^2 \, dy < c^2
\]
for suitably large \( r \). The axiom (HAP) is more subtle. In fact, it holds for bounded spectrum \( \Omega \), where it is a consequence of the Plancherel–Polya inequality for entire functions of exponential growth [50]. However, one can show that (HAP) fails for unbounded spectra, therefore
Theorem 2.2 is not directly applicable. In this case one applies the sampling part of the density theorem to the subspace $B_{\Omega} \cap B_{\delta}(0) \subseteq B_{\Omega}$ and then takes the limit $R \to \infty$. For the interpolation part, we do not need axiom (HAP)—cf. Remark 4.3—so there is no difficulty for unbounded spectrum $\Omega$, see [37] for the details.

To summarize, Theorem 2.2 implies Landau’s fundamental density theorem for bandlimited functions. The geometric properties and all but one property of the kernel are obvious, but the homogeneous approximation property requires some mathematical arguments.

### 5.2. Functions of variable bandwidth

Next we consider the spectral subspaces of the Schrödinger operator $D_q f = -\frac{1}{4\pi^2} f'' + q f$ in dimension 1 with a compactly supported potential $q \in C^2$. Let $\Omega \subseteq \mathbb{R}^+$ be a bounded set and let $\text{PW}_\Omega(D_q)$ be the spectral subspace corresponding to spectrum $\Omega$. If $q \equiv 0$, then $D_0 = -\frac{1}{4\pi^2} \frac{d^2}{dx^2}$ is diagonalized by the Fourier transform $\mathcal{F}$ so that $\mathcal{F} D_0 \mathcal{F}^{-1} f(\xi) = \xi^2 \hat{f}(\xi)$ is the operator of multiplication by $\xi^2$. For the spectral subspace $\text{PW}_\Omega(D_0)$ only the spectral values $\xi^2 \in \Omega$ are relevant, therefore $\text{PW}_\Omega(D_0) = \{ f \in L^2(\mathbb{R}) : \text{supp} \hat{f} \subseteq \Omega^{1/2} \} = B_{\Omega^{1/2}}$ is the Paley–Wiener space of bandlimited functions with spectrum in $\Omega^{1/2} = \{ \xi \in \mathbb{R} : \xi^2 \in \Omega \}$. One can show that $\text{PW}_\Omega(D_q)$ is a reproducing kernel Hilbert space.

If $q \not\equiv 0$, then we may consider $\text{PW}_\Omega(D_q)$ as a perturbation of the Paley–Wiener space. It is therefore natural to expect that the same density conditions for sampling and interpolation also hold for $\text{PW}_\Omega(D_q)$. Indeed, in [24] we proved the following result.

**Theorem 5.1.** Assume that $\Omega \subseteq \mathbb{R}^+$ is a bounded set with positive (Lebesgue) measure.

(i) If $\Lambda$ is a set of sampling for $\text{PW}_\Omega(D_q)$, then $D^- (\Lambda) \geq \mu(\Omega^{1/2})$.

(ii) If $\Lambda$ is a set of interpolation for $\text{PW}_\Omega(D_q)$, then $D^+ (\Lambda) \leq \mu(\Omega^{1/2})$.

While this result is expected, its proof is surprisingly difficult. In contrast to the Paley–Wiener space $\text{PW}_\Omega(D_0)$, the reproducing kernel for $\text{PW}_\Omega(D_q)$ is not known explicitly. To derive Theorem 5.1, we had to use the fine details of the scattering theory for one-dimensional Schrödinger operators for the verification of the kernel axioms (WL) and (HAP) and for the computation of the averaged trace of the kernel.

Thus for this example the main efforts in [24] were devoted to deriving suitable kernel estimates.

**Remark 5.2.** (i) For $\text{PW}_\Omega(D_q)$ one can also derive sufficient conditions for sampling. See [40] for a qualitative sampling theorem and [24] for an explicit almost optimal sampling theorem.

(ii) In [24] we treated a unitarily equivalent model of the Paley–Wiener space and studied a new concept of variable bandwidth. In that case the density results are formulated differently and also involve a different geometry.

### 5.3. Sampling in locally compact groups

Let $G$ be a locally compact group with Haar measure $d\mu = dx$. We make the following additional assumptions.

(i) $G$ is compactly generated, that is, there exists a symmetric neighborhood $U = U^{-1}$ of $e$ with compact closure such that $G = \bigcup_{n=0}^{\infty} U^n$. The corresponding metric on $G$, the so-called word metric, is defined as

$$d(x,y) := \min\{ n \in \mathbb{N}_0 : x^{-1}y \in U^n \}, \quad x,y \in G.$$
It is clearly left-invariant, and the balls are compact sets, in particular Borel sets of finite measure.

(ii) \( \mathcal{G} \) has polynomial growth, that is, there exist constants \( C, D > 0 \) such that
\[
\mu(U^n) \leq C n^D, \quad n \in \mathbb{N}.
\]

Under these assumptions the word metric possesses the weak annular decay property. In fact, Tessera \cite[Corollary 10]{48} showed that polynomial growth implies the annular decay property. Thus \((\mathcal{G}, d, \mu)\) satisfies all geometric axioms. (See also \cite[11]{8}.)

Now let \( \pi \) be an irreducible, unitary, square-integrable representation on a Hilbert space \( \mathcal{H} \). The orthogonality relations for square-integrable representations \cite{18} allow us to identify \( \mathcal{H} \) with a reproducing kernel Hilbert space. Precisely, fix a non-zero \( g \in \mathcal{H} \) with normalization \( \|g\| = 1 \) and consider the map
\[
\mathcal{C} : \mathcal{H} \to L^2(\mathcal{G}), \quad Cf(x) = \langle f, \pi(x)g \rangle, \quad x \in \mathcal{G}.
\]

The orthogonality relations then imply that
\[
\langle Cf, Ch \rangle_{L^2(\mathcal{G})} = d_\pi^{-1} \langle f, h \rangle_{\mathcal{H}},
\]
where the constant \( d_\pi \) is the so-called formal dimension of \( \pi \). Consequently, \( \mathcal{C} \) is a multiple of an isometry, and we can identify the representation space \( \mathcal{H} \) with the subspace \( \tilde{\mathcal{H}} = \mathcal{C} \mathcal{H} \) of \( L^2(\mathcal{G}) \). Now choosing \( h = \pi(x)g \) for \( x \in \mathcal{G} \) in \((44)\), we obtain that
\[
Cf(x) = \langle f, \pi(x)g \rangle_{\mathcal{H}} = d_\pi \langle Cf, \mathcal{C}(\pi(x)g) \rangle_{L^2(\mathcal{G})} = d_\pi \int_{\mathcal{G}} \mathcal{C}f(y) \overline{\pi(x)g, \pi(y)g} \, dy.
\]

This identity says that \( \tilde{\mathcal{H}} \) is a reproducing kernel Hilbert space with kernel
\[
k(x, y) = d_\pi \overline{\pi(x)g, \pi(y)g} = d_\pi \langle g, \pi(y^{-1}x)g \rangle.
\]
Consequently, \( k(x, x) = d_\pi \) is constant, and axiom \((D)\) is satisfied trivially. The computation of the averaged trace is a banality and yields
\[
\text{tr}^\pm(k) = \frac{1}{\mu(B_r(x))} \int_{B_r(x)} k(y, y) \, d\mu(y) = d_\pi.
\]

Moreover, since \( x \to \langle g, \pi(x)g \rangle \) is in \( L^2(\mathcal{G}) \), the weak localization (WL) is also satisfied. Again, the homogeneous approximation property (HAP) is the least obvious property and requires some work. Let \( \mathbf{B} \) consist of all vectors \( g \in \mathcal{H} \) of the form \( g = \int_{\mathcal{G}} \eta(x) \pi(x)g_0 \, d\mu(x) \) for some \( g_0 \in \mathcal{H} \) and \( \eta \) a compactly supported continuous function on \( \mathcal{G} \). If \( g \in \mathbf{B} \) and \( \Lambda \subseteq \mathcal{G} \) is an arbitrary relatively separated set, then the set of reproducing kernels \( \{ \langle g, \pi(\lambda^{-1}.)g \rangle : \lambda \in \Lambda \} \) satisfies axiom (HAP) by an observation in \cite{23}.

To formulate Theorem 2.2 for this particular example, we finally note that \( \Lambda \subseteq \mathcal{G} \) is a set of sampling (set of interpolation) for \( \tilde{\mathcal{H}} \) if and only if \( \{ \pi(\lambda)g : \lambda \in \Lambda \} \) is a frame (Riesz sequence) for \( \mathcal{H} \). Frames of this form are often called coherent frames or discrete subsets of coherent states. Theorem 2.2 yields the following density result for coherent frames.

**Theorem 5.3.** Let \( \mathcal{G} \) be a compactly generated, locally compact group with polynomial growth, and let \( \pi \) be an irreducible, unitary, square-integrable representation on a Hilbert space \( \mathcal{H} \).

(i) If \( \{ \pi(\lambda)g : \lambda \in \Lambda \} \) is a frame for \( \mathcal{H} \) for \( g \in \mathbf{B} \), then \( D^-((\Lambda)) \geq d_\pi \).
(ii) If \( \{ \pi(\lambda)g : \lambda \in \Lambda \} \) is a Riesz sequence in \( \mathcal{H} \), then \( D^+(\Lambda) \leq d_\pi \).

This result seems to be new. For square-integrable representations of groups of polynomial growth it provides a critical density that separates frames from Riesz sequences. For concrete
representations, for example, the Schrödinger representation of the Heisenberg group Theorem 5.3 has been derived many times in the context of Gabor analysis [26]. For homogeneous (nilpotent) groups it has been proved in the thesis of Höfler [27] by using the techniques of Ramanathan and Steger [42].

Let us mention that the construction of coherent frames associated to irreducible representations was first studied systematically in coorbit theory, see [16, 22]. If Λ ⊆ G is ‘sufficiently dense’, then \( \{ \pi(\lambda)g : \lambda \in \Lambda \} \) is a frame for \( \mathcal{H} \). Theorem 5.3 complements the existence of such frames by a critical density.

Theorem 2.2 also yields several new density results about sampling and interpolation in reproducing kernel Hilbert spaces that are invariant under a group action. As the full exploitation of Theorem 2.2 is beyond the scope of this section, we will come back to it in further work.

5.4. Complex analysis

Finally, we deal with sampling and interpolation in weighted spaces of analytic functions. We will partially rederive Lindholm’s result [33] from Theorem 2.2.

Let \( \phi \) be a plurisubharmonic function on \( \mathbb{C}^n \) which is 2-homogeneous and \( C^2 \) on \( \mathbb{C}^n \setminus \{0\} \). We also assume that there exist \( A, B > 0 \) such that

\[
A \cdot \text{Id}_n \leq (\partial_j \partial_k \phi(z))_{j,k=1,...,n} \leq B \cdot \text{Id}_n \tag{45}
\]

for all \( z \neq 0 \), in the sense of positive definite matrices. It follows that \( A^n \leq \det(\partial_j \partial_k \phi)_{j,k} \leq B^n \) on \( \mathbb{C}^n \setminus \{0\} \). Note that in dimension \( n = 1 \) this condition simply means that the Laplacian \( \Delta \phi = \partial \bar{\partial} \phi \) is bounded above and below from 0.

Our main object is the Hilbert space \( \mathcal{F}_2^\phi \) of entire functions on \( \mathbb{C}^n \) defined by the norm

\[
\| f \|_{\mathcal{F}_2^\phi} = \int_{\mathbb{C}^n} |f|^2 e^{-2\phi} dm < \infty \tag{47}
\]

We observe immediately that the assumptions on the metric and the measure required in the main theorem are satisfied.

It can be shown that \( \mathcal{F}_2^\phi \) is a reproducing kernel Hilbert space. We denote the kernel by \( K = K_\phi \). For the weight \( \phi(z) = |z|^2/2 \), our normalization yields the following explicit expression for the kernel

\[
K(z, w) = \frac{2^n}{\pi^n} e^{\frac{z \cdot \bar{w} - |z|^2/2 - |w|^2/2}{4}}.
\]

It is easy to see that this kernel satisfies the axioms (D), (WL), and (HAP), therefore the Seip’s necessary density conditions [44] for sampling in Bargmann–Fock space follow (without strict inequalities) directly from Theorem 2.2.

For more general weights \( \phi \) there is no explicit formula for the kernel, but strong estimates are known. We use Lindholm’s estimates [33]. Since he works with the measure \( e^{-\phi} dm \) and
entire functions, we have to translate these results to our normalization with the measure $\mu$ and functions of the form (46). Using the observation (22), the relation between the kernel $B_\phi$ in [33] and our kernel $K_\phi$ is given by

$$K_\phi(z,w) = \frac{1}{\sqrt{\det(\partial_j \bar{\partial}_k \phi)_{jk}(z) \det(\partial_j \bar{\partial}_k \phi)_{jk}(w)}} B_\phi(z,w) e^{-\phi(z)-\phi(w)}.$$  

Translated to our notation, Lindholm [33] proved the following facts about $K_\phi$.

(i) There exist constants $C,T > 0$ depending on $A$ and $B$ in (45), such that for all $k > 0$ holds the decay estimate

$$K_{k^2 \phi}(z,w) \leq C e^{-kT|z-w|}, \quad (48)$$

(ii) On the diagonal the kernel satisfies the limit relation

$$\lim_{k \to \infty} K_{k^2 \phi}(z,z) = \frac{2^n}{\pi^n}, \quad (49)$$

with uniform convergence on $\mathbb{C}^n \setminus B_\tau(0)$ for arbitrary $\tau > 0$.

In addition, the 2-homogeneity and a simple change of variables imply that

$$K_{k^2 \phi}(z,w) = K_\phi(kz,kw). \quad (50)$$

The axioms (WL) and (HAP) follow immediately from the off-diagonal decay (48) of the kernel $K_\phi$, likewise $K_\phi(z,z)$ is bounded. The lower bound in Axiom (D) (6) can now be verified in the following way. Note first that since $g = (\det(\partial_j \bar{\partial}_k \phi)_{jk})^{-1/2} e^{-\phi} \in A^2_\phi$, it follows that $K(z,z) \neq 0$, for all $z \in \mathbb{C}^n$. In addition, by (49) and (50), there exist $c, \tau, R > 0$ such that $K_{R^2 \phi}(z,z) = K_\phi(R \cdot z/R, R \cdot z/R) > c$ for any $|z| > R \tau$. By continuity and the fact that $K(z,z) \neq 0$ for all $z$, we also have $K(z,z) > c_2$ for some $c_2 > 0$ for $|z| \leq R \tau$.

Thus the geometry and the kernel satisfy all required hypotheses of Section 2.1. Thus Theorem 2.2 is applicable and yields a critical density that separates sampling from interpolation. It remains to compute this critical density.

**Lemma 5.4.** If the weight $\phi$ is plurisubharmonic, 2-homogeneous, and satisfies (45), then

$$\text{tr}^+(K_\phi) = \text{tr}^-(K_\phi) = \frac{2^n}{\pi^n}. \quad (51)$$

**Proof.** We use the homogeneity (50) and $(\partial_j \bar{\partial}_k \phi)(rz) = \partial_j \bar{\partial}_k \phi(z)$ for all $r > 0$. Then

$$\sup_{x \in \mathbb{C}^n} \int_{B_r(x)} \frac{1}{\det(\partial_j \bar{\partial}_k \phi)_{jk}(z) \det(\partial_j \bar{\partial}_k \phi)_{jk}(w)} K_\phi(z,z) d\mu(z)$$

$$= \sup_{y \in \mathbb{C}^n} \int_{B_1(y)} \frac{1}{\det(\partial_j \bar{\partial}_k \phi)_{jk}(z) \det(\partial_j \bar{\partial}_k \phi)_{jk}(w)} K_{R^2 \phi}(w,w) d\mu(w)$$

$$= \sup_{y \in \mathbb{C}^n} \int_{B_1(y)} \frac{1}{\det(\partial_j \bar{\partial}_k \phi)_{jk}(z) \det(\partial_j \bar{\partial}_k \phi)_{jk}(w)} K_{R^2 \phi}(w,w) d\mu(w). \quad (52)$$
Now, we use the fact that $K_{\tau^2\phi}(w, w)$ converges to $2^n/\pi^n$ uniformly outside any ball $B_{\tau}(0)$, $\tau > 0$. If $B_{\tau}(y)$ contains the origin, we remove a small neighborhood of 0, otherwise we use (49) directly. Given $\epsilon > 0$, it follows that

$$\lim_{r\to\infty} \sup_{x \in \mathbb{C}^n} \left| \frac{1}{\det(\partial_j \partial_k \phi)_{jk} dm} \int_{B_r(x)} K_{\tau^2\phi}(z, z) d\mu(z) - \frac{2^n}{\pi^n} \right| \leq \epsilon + o(1)$$

as $r \to \infty$. Therefore,

$$\lim_{r\to\infty} \sup_{x \in \mathbb{C}^n} \left| \frac{1}{\det(\partial_j \partial_k \phi)_{jk} dm} \int_{B_r(x)} K_{\phi}(z, z) dm(z) - \frac{2^n}{\pi^n} \right| \leq \epsilon$$

for all $\epsilon > 0$, which means that

$$\text{tr}^+(K_{\phi}) = \lim_{r\to\infty} \sup_{x \in \mathbb{C}^n} \left| \frac{1}{\det(\partial_j \partial_k \phi)_{jk} dm} \int_{B_r(x)} K_{\phi}(z, z) dm(z) = \frac{2^n}{\pi^n} \right.$$

Likewise $\text{tr}^-(K_{\phi}) = 2^n/\pi^n$. \hfill \Box

**Theorem 5.5.** Assume that $\phi$ is plurisubharmonic, 2-homogeneous, and satisfies (45).

(i) If $\Lambda \subseteq \mathbb{C}^d$ is a set of sampling for $A^2_{\phi}$, then $D^-(\Lambda) \geq 2^n/\pi^n$.

(ii) If $\Lambda \subseteq \mathbb{C}^d$ is a set of interpolation for $A^2_{\phi}$, then $D^+(\Lambda) \leq 2^n/\pi^n$.

**Remark 5.6.** For generalized Fock spaces in one complex variable, Ortega-Cerdà and Seip [38] and Marco, Massaneda, Ortega-Cerdà [36] proved a density theorem for non-homogeneous weights as well. Although Theorem 2.2 applies, we are (not yet) able to recover their explicit result. This would require to derive a version of Lemma 5.4 for non-homogeneous doubling weights.

5.5. Density of abstract frames

Finally, we note certain connections with the density theory for abstract frames [4–6].

Let $\mathcal{H}$ be a separable Hilbert space, $(X, d)$ a countable metric space with counting measure $\mu$, and $\mathcal{F} = \{f_x \in \mathcal{H} \}$ a frame for $\mathcal{H}$, that is, there exist $A, B > 0$ such that

$$A\|f\|^2 \leq \sum_{x \in X} |\langle f, f_x \rangle|^2 \leq B\|f\|^2 \quad \forall f \in \mathcal{H}. \quad (53)$$

Using the coefficient operator $\mathcal{C} : \mathcal{H} \to \ell^2(X)$

$$\mathcal{C}f(x) = \langle f, f_x \rangle \quad x \in X, \quad (54)$$

we can identify the abstract Hilbert space $\mathcal{H}$ with the subspace of functions $\mathcal{C}f$ of $\ell^2(X)$. By the frame inequalities (53) $\mathcal{C}$ is one-to-one with closed range in $\ell^2(X)$, which we call $\mathcal{H} = \mathcal{C}\mathcal{H} \subseteq \ell^2(X)$.

Let $\{f_x : x \in X\}$ be the (canonical) dual frame of $\mathcal{H}$, then every $f \in \mathcal{H}$ possesses the frame expansion $f = \sum_{y \in X} \langle f, f_y \rangle f_y$, and consequently

$$\mathcal{C}f(x) = \sum_{y \in X} \langle f, f_y \rangle \langle f_y, f_x \rangle = \sum_{y \in X} \mathcal{C}f(y) \langle f_y, f_x \rangle,$$

this means that $\mathcal{H}$ is a reproducing kernel subspace of $\ell^2(X)$ with kernel

$$k(x, y) = \langle f_y, f_x \rangle.$$
The two properties (WL) and (HAP) for the pair \((\mathcal{X}, \mathcal{H})\) that are equivalent to what in [4] is called \(\ell^2\)-localization of the frames \(\mathcal{F}\) and \(\tilde{\mathcal{F}}\). Furthermore, the (lower) averaged trace of this kernel is

\[
tr^- (k) = \lim_{r \to \infty} \inf_{x \in \mathcal{X}} \inf_{z \in X} \frac{1}{\# B_r(x)} \sum_{y \in B_r(x)} \langle \tilde{f}_y, f_y \rangle.
\]

This quantity correspond exactly to the (lower) frame measure of \(\mathcal{F}\) in [4, 6].

Besides these technical similarities, Theorem 2.2 is not formally comparable to the results in [4–6]. The theory of Balan, Casazza, Heil, and Landau in [4–6] compares two abstract frames, and derives an equality relating density and measure. By contrast, Theorem 2.2 compares a frame of reproducing kernels to a possibly continuous resolution of the identity.

5.6. More on Axiom (WAD) — The standard Bergman space on the upper-half plane

Let \(X = \{z \in \mathbb{C} : \text{Im}(z) > 0\}\) with the hyperbolic distance

\[
d(z, w) = 2 \tanh^{-1} \left( \frac{|z - w|}{|z - w|} \right),
\]

and measure \(d\mu(z) = \frac{1}{2} \text{Im}(z)^{-2} dA(z)\), where \(dA(z)\) denotes the Lebesgue measure. We consider the RKHS of functions

\[
\mathcal{H} = \{\text{Im}(z)f(z), \text{ with } f : X \to \mathbb{C} \text{ analytic} \} \cap L^2(X, \mu).
\]

One can readily verify that the measure of \(B_R(0)\) grows exponentially in \(R\) and that the weak annular decay property does not hold. Hence, Theorem 2.2 is not applicable in this setting. Nevertheless, with the appropriate notion of density introduced by Seip [45], necessary and sufficient conditions for sampling and interpolation do hold for \(\mathcal{H}\). However, this is not the usual Beurling density.

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