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# Geometric Learning with Positively Decomposable Kernels 

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#### Abstract

Kernel methods are powerful tools in machine learning. Classical kernel methods are based on positive-definite kernels, which map data spaces into reproducing kernel Hilbert spaces (RKHS). For non-Euclidean data spaces, positive-definite kernels are difficult to come by. In this case, we propose the use of reproducing kernel Krein space (RKKS) based methods, which require only kernels that admit a positive decomposition. We show that one does not need to access this decomposition in order to learn in RKKS. We then investigate the conditions under which a kernel is positively decomposable. We show that invariant kernels admit a positive decomposition on homogeneous spaces under tractable regularity assumptions. This makes them much easier to construct than positive-definite kernels, providing a route for learning with kernels for non-Euclidean data. By the same token, this provides theoretical foundations for RKKS-based methods in general.


Keywords: kernel methods, geometric learning, Krein spaces, non-positive kernels, invariant kernels, Gaussian kernel, homogeneous spaces, symmetric spaces

## 1 Introduction

Kernel methods have proved to be successful in machine learning. Examples of algorithms involving kernels include support vector machines (SVM) (Cristianini and Ricci (2008)), kernel principal component analysis (Schölkopf et al. (1997)), and maximum mean discrepancy (Gretton et al. (2012)). Classically, kernel methods rely on a positive-definite (PD) kernel on the data space. Such a PD kernel gives an embedding of the data space into a reproducing kernel Hilbert space (RKHS), which is a space of functions over the data space. Learning problems are then often phrased as learning a function in the RKHS, which can be done effectively thanks to the Representer Theorem (Schölkopf et al. (2001)).

Kernel methods on Euclidean data spaces have been thoroughly studied and numerous families of PD kernels have been proposed in Euclidean spaces, many of which have universal approximation properties (Steinwart (2001); Micchelli et al. (2006)). However, in many

[^0]applications it is important to capture the geometry of the data, and view it as lying on a non-Euclidean space, such as a manifold. Even though there has been considerable recent interest in constructing PD kernels on manifolds (Jayasumana et al. (2013); Feragen et al. (2014); Borovitskiy et al. (2022); Azangulov et al. (2023a b)), it has proved difficult to find closed form PD kernels on general geometries.

Ong et al. (2004) proposed a different class of kernel algorithms, that do not require positive-definiteness of the kernel. Instead, they require the existence of a positive (PD) decomposition for the kernel, meaning that it can be written as a difference of PD kernels. Such a kernel embeds the data space into a reproducing kernel Krein space (RKKS), consisting of functions on the data space. As in the RKHS case, this allows for the solution of learning problems from data (Schleif and Tino (2015); Loosli et al. (2016); Oglic and Gaertner (2018); Liu et al. (2021a b)) via, for instance, an adapted representer theorem.

We are interested in the problem of characterizing kernels which admit a PD decomposition, both on Euclidean and non-Euclidean data spaces. Providing such a characterization promises to solve several open problems in the literature. More specifically, it achieves the following.

1. It justifies the use of non-PD kernels for RKHS-based methods. In applied areas, nonPD kernels have sometimes been used when restricted to a data set for which the Gram matrix corresponding to this kernel is PD. This has, for example, been done with the geodesic Gaussian kernel on non-Euclidean geometries (Calinon (2020); Jousse et al. (2021)), which is known not to be PD in general (Feragen et al. (2014); Da Costa et al. (2023); Li (2023)). It seems that many RKHS-based methods (such as support vector machines (Cristianini and Ricci (2008))) may be generalized to RKKS-based methods (Loosli et al. (2016)), such that the solution to both learning problems are the same, when restricted to a finite data set on which the kernel is PD. Therefore applying an RKHS-based method to a non-PD kernel, which is PD on a given finite data set, can be justified by the existence of a PD decomposition for the kernel.
2. It justifies the use of RKKS-based methods. Analogous to how a kernel may be PD on a finite data set but not on the whole space, a kernel may be PD decomposable on a finite data set but not on the whole space. Except that in this case, this subtlety is even easier to ignore, as kernels are always PD decomposable on finite data sets. However, if a kernel does not admit a PD decomposition on the whole data space, then the solution to the learning problem has no guarantees to generalize to unseen data. RKKS-based methods have so far largely ignored the issue of checking PD decomposability of the kernels used. Only linear combinations of PD kernels where truly known to admit a PD decomposition. Recently, a theoretical advance was made by Liu et al. (2021a) by observing that invariant kernels on $X=\mathbb{R}^{n}$ admit a PD decomposition if and only if they are the inverse Fourier transform of a finite signed measure. We will generalize this result to many non-Euclidean data spaces $X$.
3. It motivates the use of RKKS-based methods for non-Euclidean data spaces. In the Euclidean case $X=\mathbb{R}^{n}$, many families of kernels are known to be PD. RKHS-based methods are, therefore, usually sufficient in this case. In contrast, for non-Euclidean data spaces $X$, PD kernels can be challenging to construct. For example, it has
been shown that the generalization of some of the most widely used kernels in $\mathbb{R}^{n}$, such as the Gaussian kernel, are often not PD on non-Euclidean geometries (Feragen et al. (2014); Da Costa et al. (2023); Li (2023)). Positive-definiteness of a kernel is a stringent condition, and heavily depends on the specific geometry of the data space in question. This makes it difficult to describe general families of kernels that are PD on general classes of geometries. Indeed, closed form PD kernels have only been found on a handful of geometries (Feragen et al. (2014)). In contrast, we will see that PD decomposability is a much weaker condition to impose on a kernel, and it will be enough to require some symmetry and regularity assumptions. Therefore, in non-Euclidean settings, RKKS-based methods become particularly compelling.

Paper organization and contributions: This paper aims to provide verifiable sufficient conditions for kernels to admit a positive decomposition, thereby justifying their use for RKKS learning. We pay particular attention to invariant kernels on locally compact groups $G$ and their coset spaces $G / H$, and then look into the special case where $G$ is a Lie group, and $G / H$ is a homogeneous space.

We start by briefly reviewing the theory of PD kernels and how they permit RKHS learning in Section 2.1. We then describe the analogous theory of PD decomposable kernels and RKKS learning in Section 2.2. In particular, we provide a new rigorous proof to a general representer theorem for RKKS learning. Crucially, we show that, when the regularizer of the learning problem is linear in the squared indefinite inner product, one does not need access to the PD decomposition of the kernel in order to apply the representer theorem; one only needs to know that it exists.

In Section 3, we focus on invariant kernels on homogeneous spaces of locally compact groups and prove the correspondence between ( $\mathrm{PD} / \mathrm{PD}$ decomposable) invariant kernels on $G / H$ and (PD/PD decomposable) Hermitian functions on the double coset space $H \backslash G / H$. When $H=\{e\}$, this allows us to relate PD decomposable invariant kernels on $G$ to realvalued functions in the Fourier-Stieltjes algebra $B_{\mathbb{C}}(G)$, which in turn allows us to leverage results from harmonic analysis.

In Section 4, we study the case where $G$ is commutative, where the situation simplifies considerably. In particular, we review known sufficient conditions for functions to belong to the spaces $B_{\mathbb{C}}\left(\mathbb{R}^{n}\right)$ and $B_{\mathbb{C}}\left(S^{1}\right)$. As an example, we show that the Gaussian kernel, despite failing to be PD for instance on the torus $\mathbb{T}^{n}$, has a PD decomposition on any Abelian Lie group.

In Section 5, we move to the case where $G$ is not assumed to be commutative. In this case the literature has focused on the space of functions $B_{\mathbb{C}}(G)$, and not the more general $B_{\mathbb{C}}(H \backslash G / H)$. To fill this gap, we show in Section 5.1, using the language of representation theory, that any function that is invariant on (double) cosets and is PD decomposable, has a PD decomposition into functions that are invariant on (double) cosets, i.e. $B_{\mathbb{C}}(G) \cap$ $C(G / H)=B_{\mathbb{C}}(G / H)$ and $B_{\mathbb{C}}(G) \cap C(H \backslash G / H)=B_{\mathbb{C}}(H \backslash G / H)$. This allows us to show in Section 5.2 that on a homogeneous space $G / H$, with $G$ a unimodular Lie group and $H$ a compact Lie subgroup, smooth functions whose derivatives decay appropriately at infinity admit PD decompositions. As an example, we then show that the Gaussian kernel has a PD decomposition on non-compact symmetric spaces.

## 2 Kernels and Representer Theorems

### 2.1 Positive-definite kernels

Definition $1 A$ kernel on a set $X$ is a Hermitian map $k: X \times X \rightarrow \mathbb{C}$. $k$ is said to be positive-definite $(P D)$ if for all $N \in \mathbb{N}, x_{1}, \ldots, x_{N} \in X$ and all $c_{1}, \ldots, c_{N} \in \mathbb{C}$,

$$
\sum_{i=1}^{N} \sum_{j=1}^{N} \overline{c_{i}} c_{j} k\left(x_{i}, x_{j}\right) \geq 0
$$

i.e. the matrix $\left(k\left(x_{i}, x_{j}\right)\right)_{i, j}$, which we call the Gram matrix of $x_{1}, \ldots, x_{N}$, is Hermitian positive semi-definite.

In practice, we are interested in real-valued kernels $k: X \times X \rightarrow \mathbb{R}$, but allowing kernels to take complex values will be convenient in the following theoretical work.

Remark 2 In the literature, the term 'kernel' is sometimes used to mean PD kernel and occasionally even the Hermitian assumption is dropped. We refer to (Berg et al., 1984, Chapter 3) for the general theory of PD kernels.

In machine learning, a key reason for the importance of PD kernels is the following theorem (see (Paulsen and Raghupathi, 2016, Theorem 3.16)) that guarantees that PD kernels have a natural associated reproducing kernel Hilbert space (RKHS):

Theorem 3 Let $k$ be a PD kernel on a set $X$. Then there is a complex Hilbert space $\mathcal{H}$, which we call the reproducing kernel Hilbert space (RKHS) associated with $k$, consisting of complex-valued functions on $X$, with a map

$$
\begin{align*}
\Phi: X & \rightarrow \mathcal{H} \\
x & \rightarrow k(x, \cdot) \tag{1}
\end{align*}
$$

such that

1. $\langle\Phi(x), \Phi(y)\rangle_{\mathcal{H}}=k(x, y)$ for all $x, y \in X$,
2. $\overline{\operatorname{span}(\Phi(X))}=\mathcal{H}$.

Any map $\Phi$ that satisfies the equality $\langle\Phi(x), \Phi(y)\rangle_{\mathcal{H}}=k(x, y)$ in the first part of the theorem is called a feature map of the RKHS $\mathcal{H}$. The functions in the image of the feature map $\Phi$ in (1) are called kernel sections of $k$. In practice, kernels will be real-valued, so we will usually consider the induced real RKHS.

Suppose we are trying to learn a real or complex-valued function $f$ over $X$ from a finite set of observations. One can try to find a good approximation for this function in the RKHS associated with a particular kernel map $k$. This is particularly pertinent when the kernel in question has universality properties (Steinwart (2001); Micchelli et al. (2006)), that is, when its sections are dense, say, in the set of continuous functions in $X$. The representer theorem in RKHS (Schölkopf et al. (2001)) tells us that, for a general class of learning problems, the best approximation $f^{*}$ in the RKHS, given some observations $D=\left\{\left(x_{i}, y_{i}\right)_{i=1}^{N}\right\} \subset(X \times \mathbb{R})^{N}$ of the function mapping the $x_{i}$ to the $y_{i}$, is a finite linear combination of the kernel sections $k\left(x_{i}, \cdot\right)$. For the sake of completeness, we review a version of this theorem.

Theorem 4 (Representer Theorem in RKHS) Let $k$ be a real-valued PD kernel on a set $X$ with $\mathcal{H}$ its associated real $R K H S$, and let

1. $D=\left\{\left(x_{1}, y_{1}\right), \ldots,\left(x_{N}, y_{N}\right)\right\} \subset(X \times \mathbb{R})^{N}$ be a finite data set,
2. $g:[0, \infty) \rightarrow \mathbb{R}$ be a strictly increasing function,
3. $L: \mathcal{H} \times(X \times \mathbb{R})^{N} \rightarrow \mathbb{R}$ be a loss functional, determined exclusively through function evaluations, i.e. if $f, g \in \mathcal{H}$ are such that $\left.f\right|_{D}=\left.g\right|_{D}$, then $L(f, D)=L(g, D)$,
4. $\Omega(D) \subset \mathcal{H}$ be a feasible set, determined exclusively through function evaluations, i.e. if $f, g \in \mathcal{H}$ are such that $\left.f\right|_{D}=\left.g\right|_{D}$, then $f \in \Omega(D)$ if and only if $g \in \Omega(D)$.

Then if the optimization problem

$$
\begin{equation*}
\underset{f \in \mathcal{H}}{\operatorname{minimize}} L(f, D)+g\left(\|f\|_{\mathcal{H}}\right) \quad \text { with respect to } \quad f \in \Omega(D) \tag{2}
\end{equation*}
$$

admits a unique solution $f^{*}$, we have

$$
f^{*}=\sum_{i=1}^{N} \alpha_{i} k\left(x_{i}, \cdot\right)
$$

for some $\alpha_{1}, \ldots, \alpha_{N} \in \mathbb{R}$.
The problem is therefore reduced to learning the coefficients $\alpha_{i}$. We have thus transformed an infinite-dimensional problem (learning a function) into a finite-dimensional one.

Example 5 Support vector machines (SVM) (Vapnik and Chervonenkis (1968); Vapnik (1998); Cristianini and Ricci (2008)) are a class of learning algorithms that aim at the binary classification of data. Given a set $X$, a PD kernel $k$ on $X, \mathcal{H}$ the RKHS associated to $k$, and a finite data set $D=\left\{\left(x_{1}, y_{1}\right), \ldots,\left(x_{N}, y_{N}\right)\right\} \subset(X \times\{-1,1\})^{N}$, the learning problem solved by binary SVM can be written as

$$
\begin{equation*}
\underset{f \in \mathcal{H}, b \in \mathbb{R}}{\operatorname{minimize}}\|f\|_{\mathcal{H}}^{2} \quad \text { with respect to } \quad \sum_{i=1}^{N} \max \left(0,1-y_{i}\left(f\left(x_{i}\right)+b\right)\right) \leq \tau \tag{3}
\end{equation*}
$$

for some fixed parameter $\tau>0$. The representer theorem tells us that any solution $(f, b)$ to this problem has $f=\sum_{i=1}^{N} \alpha_{i} k\left(x_{i}, \cdot\right)$ for some $\alpha_{i} \in \mathbb{R}$.

Note that to be precise, we are applying here the parametric extension to the representer theorem (Schölkopf et al., 2001, Theorem 2), since the minimization is over $f$ and $b$, and not solely over $f$ as in Theorem 4.

### 2.2 Positively decomposable kernels

A natural question that can be posed is whether the assumption on the kernel $k$ being positive-definite is necessary? That is, can we still obtain a result analogous to Theorem 4 while dropping this assumption? It turns out that we can. For this we need to define the notion of a Krein space. We refer to Bognár (1974) for a thorough presentation of Krein spaces, and to Schwartz (1964) and Alpay (1991) for the study of their reproducing kernels. This last reference contains a good summary of Schwartz's contributions.

Definition 6 An indefinite inner product on a real vector space $\mathcal{K}$ is a Hermitian bilinear $\operatorname{map}\langle\cdot, \cdot\rangle: X \times X \rightarrow \mathbb{R}$ which is non-degenerate, that is, for all $f \in \mathcal{K}$,

$$
\langle f, f\rangle=0 \quad \text { implies that } \quad f=0
$$

A complex vector space $\mathcal{K}$ equipped with an indefinite inner product $\langle\cdot, \cdot\rangle$ is called a Krein space if it can be written as the algebraic direct sum

$$
\begin{equation*}
\mathcal{K}=\mathcal{H}_{+} \oplus \mathcal{H}_{-} \tag{4}
\end{equation*}
$$

such that

1. $\mathcal{H}_{+}$equipped with $\langle\cdot, \cdot\rangle_{+}:=\langle\cdot, \cdot\rangle$ is a Hilbert space,
2. $\mathcal{H}_{-}$equipped with $\langle\cdot, \cdot\rangle_{-}:=-\langle\cdot, \cdot\rangle$ is a Hilbert space,
3. $\left\langle f_{+}, f_{-}\right\rangle=0$ for all $f_{+} \in \mathcal{H}_{+}, f_{-} \in \mathcal{H}_{-}$.

In other words, there are complete inner products $\langle\cdot, \cdot\rangle_{+},\langle\cdot, \cdot\rangle_{-}$on $\mathcal{H}_{+}, \mathcal{H}_{-}$respectively such that

$$
\langle f, g\rangle=\left\langle f_{+}, g_{+}\right\rangle_{+}-\left\langle f_{-}, g_{-}\right\rangle_{-}
$$

for all $f=f_{+}+f_{-}, g=g_{+}+g_{-}$with $f, g \in \mathcal{K}, f_{+}, g_{+} \in \mathcal{H}_{+}, f_{-}, g_{-} \in \mathcal{H}_{-}$.
$\mathcal{K}$ is equipped with the product topology on $\mathcal{H}_{+} \times \mathcal{H}_{-}$, which can be shown to be independent of the choice of decomposition (4).

Definition 7 Let $k$ be kernel on $X$. Then $k$ is said to have a positive (PD) decomposition if it can be written as

$$
k=k_{+}-k_{-}
$$

where $k_{+}$and $k_{-}$are PD kernels.
Remark 8 PD decompositions are also sometimes called fundamental decompositions (Bognár (1974)), or Kolmogorov decompositions (Mary (2003)).

With this notion, we obtain an analogous result to Theorem 3 (see (Alpay, 1991, Theorem 2.1)):

Theorem 9 Let $k$ be a PD decomposable kernel on a set $X$. Then there is a complex Krein space $\mathcal{K}$, which we call the reproducing kernel Krein space ( $R K K S$ ) associated to $k$, consisting of complex-valued functions over $X$, with a map

$$
\begin{aligned}
\Phi: X & \rightarrow \mathcal{K} \\
x & \rightarrow k(x, \cdot)
\end{aligned}
$$

such that

1. $\langle\Phi(x), \Phi(y)\rangle_{\mathcal{K}}=k(x, y)$ for all $x, y \in X$,
2. $\overline{\operatorname{span}(\Phi(X))}=\mathcal{K}$.

Remark 10 As opposed to the PD case, the RKKS associated to a PD decomposable kernel $k$ can, in some cases, be non-unique, see Schwartz (1964) and Alpay (1991).

Theorem 9 can be made into an if and only if statement: $k$ must admit a PD decomposition in order to give rise to an RKKS, in the same way that $k$ must be PD in order to give rise to an RKHS. As in the PD case, we will usually consider only the induced real RKKS of a real-valued PD decomposable kernel.

Now as in the RKHS case (Theorem 4), a representer theorem can be formulated in the RKKS context (Ong et al. (2004)) in which the indefinite nature of the inner product implies that the optimization problem (2) is replaced by a critical point problem. In the next statement, we shall be using the following terminology: let $M$ be a (potentially infinitedimensional) manifold and let $S \subset M$ be a submanifold (potentially with boundary). Now let $f: M \rightarrow \mathbb{R}$ be a differentiable function and define the stabilization problem of $f$ with respect to $S$ to be the problem of finding the critical points of the restriction $\left.f\right|_{S}$, which we denote by

$$
\underset{m \in M}{\operatorname{stabilize}} f(m) \quad \text { with respect to } S \text {. }
$$

Theorem 11 (Representer Theorem in RKKS) Let $k$ be a real-valued PD decomposable kernel on a set $X$ with $\mathcal{K}$ an associated real $R K K S$, and let

1. $D=\left\{\left(x_{1}, y_{1}\right), \ldots,\left(x_{N}, y_{N}\right)\right\} \subset(X \times \mathbb{R})^{N}$ be a finite dataset,
2. $g: \mathbb{R} \rightarrow \mathbb{R}$ be a differentiable function,
3. $L: \mathcal{K} \times(X \times \mathbb{R})^{N} \rightarrow \mathbb{R}$ be a differentiable loss functional, determined exclusively through function evaluations,
4. I: $\mathcal{K} \times(X \times \mathbb{R})^{N} \rightarrow \mathbb{R}^{m}$ be a differentiable functional setting inequality constraints, determined exclusively through function evaluations,
5. $E: \mathcal{K} \times(X \times \mathbb{R})^{N} \rightarrow \mathbb{R}^{l}$ be a differentiable functional setting equality constraints, determined exclusively through function evaluations.

Then, if the maps $E(\cdot, D): \mathcal{K} \rightarrow \mathbb{R}^{l}$ and $I(\cdot, D): \mathcal{K} \rightarrow \mathbb{R}^{m}$ are submersions and the conditions $E(f, D)=0$ and $I(f, D) \leq 0$ determine a submanifold of $\mathcal{K}$, the solutions $f^{*}$ of the stabilization problem

$$
\begin{equation*}
\underset{f \in \mathcal{K}}{\operatorname{stabilize}} L(f, D)+g\left(\langle f, f\rangle_{\mathcal{K}}\right) \quad \text { with respect to } \quad I(f, D) \leq 0 \text { and } E(f, D)=0 \tag{5}
\end{equation*}
$$

for which the generic condition $g^{\prime}\left(\left\langle f^{*}, f^{*}\right\rangle_{\mathcal{K}}\right) \neq 0$ is satisfied, have all the form

$$
\begin{equation*}
f^{*}=\sum_{i=1}^{N} \alpha_{i} k\left(x_{i}, \cdot\right), \tag{6}
\end{equation*}
$$

for constants $\alpha_{1}, \ldots, \alpha_{N} \in \mathbb{R}$. Moreover, the if $g(x)=c x$ for some $c \in \mathbb{R}$ and all $x \in \mathbb{R}$, then the coefficients $\alpha_{1}, \ldots, \alpha_{N} \in \mathbb{R}$ do not depend on the choice of indefinite inner product $\langle\cdot, \cdot\rangle_{\mathcal{K}}$ associated to $k$.

Proof The submersion assumption on the functions $E(\cdot, D): \mathcal{K} \rightarrow \mathbb{R}^{l}$ and $I(\cdot, D): \mathcal{K} \rightarrow \mathbb{R}^{m}$ and the submanifold character of the set determined by $E(f, D)=0$ and $I(f, D) \leq 0$ imply (see (Abraham et al., 1988, Corollary 3.5.29)) that for each critical point $f^{*} \in \mathcal{K}$ of the restriction of $L(\cdot, D)+g\left(\langle\cdot, \cdot\rangle_{\mathcal{K}}\right)$ to it, there exist constants $\lambda \in \mathbb{R}^{m}, \mu \in \mathbb{R}^{l}$ such that $f^{*}$ is a critical point of the corresponding Lagrange function defined by

$$
\mathcal{L}(f)=L(f, D)+g\left(\langle f, f\rangle_{\mathcal{K}}\right)+\lambda^{\top} I(f, D)+\mu^{\top} E(f, D),
$$

that is,

$$
\begin{equation*}
D \mathcal{L}\left(f^{*}\right)=0, \tag{7}
\end{equation*}
$$

where $D \mathcal{L}: \mathcal{K} \rightarrow \mathcal{K}^{*}$ denotes the Fréchet differential of $\mathcal{L}: \mathcal{K} \rightarrow \mathbb{R}$. We recall that $\lambda_{i} I_{i}\left(f^{*}, D\right)=0$ for all $i \in\{1, \ldots, m\}$, this implies that $\lambda_{i}$ is non-zero only when the corresponding inequality constraint is active, that is $I_{i}\left(f^{*}, D\right)<0$. Additionally, the dependence of the conditions on $L, I$, and $E$ on $\mathcal{K}$ exclusively through function evaluations implies the existence of three differentiable functions $\widetilde{L}, \widetilde{I}$, and $\widetilde{E}$ defined in $\mathbb{R}^{N}$ such that $L(f, D)=$ $\widetilde{L}\left(f\left(x_{1}\right), \ldots, f\left(x_{N}\right)\right), I(f, D)=\widetilde{I}\left(f\left(x_{1}\right), \ldots, f\left(x_{N}\right)\right), E(f, D)=\widetilde{E}\left(f\left(x_{1}\right), \ldots, f\left(x_{N}\right)\right)$. With these functions and recalling that $\langle k(x, \cdot), f\rangle_{\mathcal{K}}=f(x)$, for any $f \in \mathcal{K}$ and $x \in X$, we can write, for any $f, g \in \mathcal{K}$ :

$$
D \mathcal{L}(f) \cdot g=\sum_{i=1}^{N}\left(\partial_{i} \widetilde{L}(\mathbf{f}(\mathbf{x}))+\lambda^{\top} \partial_{i} \widetilde{I}(\mathbf{f}(\mathbf{x}))+\mu^{\top} \partial_{i} \widetilde{E}(\mathbf{f}(\mathbf{x}))\right)\left\langle k\left(x_{i}, \cdot\right), g\right\rangle_{\mathcal{K}}+2 g^{\prime}\left(\langle f, f\rangle_{\mathcal{K}}\right)\langle f, g\rangle_{\mathcal{K}},
$$

where $\mathbf{f}(\mathbf{x}):=\left(f\left(x_{1}\right), \ldots, f\left(x_{N}\right)\right)$. If we now evaluate the previous expression at $f=f^{*}$ and $g=k(x, \cdot)$, for any $x \in X$, the condition (7) reads

$$
\begin{aligned}
0=\sum_{i=1}^{N}\left(\partial_{i} \widetilde{L}\left(\mathbf{f}^{*}(\mathbf{x})\right)+\lambda^{\top} \partial_{i} \widetilde{I}\left(\mathbf{f}^{*}(\mathbf{x})\right)+\mu^{\top} \partial_{i} \widetilde{E}\left(\mathbf{f}^{*}(\mathbf{x})\right)\right)\langle & \left.k\left(x_{i}, \cdot\right), k(x, \cdot)\right\rangle_{\mathcal{K}} \\
& +2 g^{\prime}\left(\left\langle f^{*}, f^{*}\right\rangle_{\mathcal{K}}\right)\left\langle f^{*}, k(x, \cdot)\right\rangle_{\mathcal{K}},
\end{aligned}
$$

or, equivalently

$$
0=\sum_{i=1}^{N}\left(\partial_{i} \widetilde{L}\left(\mathbf{f}^{*}(\mathbf{x})\right)+\lambda^{\top} \partial_{i} \widetilde{I}\left(\mathbf{f}^{*}(\mathbf{x})\right)+\mu^{\top} \partial_{i} \widetilde{E}\left(\mathbf{f}^{*}(\mathbf{x})\right)\right) k\left(x_{i}, x\right)+2 g^{\prime}\left(\left\langle f^{*}, f^{*}\right\rangle_{\mathcal{K}}\right) f^{*}(x) .
$$

Now, since by hypothesis $g^{\prime}\left(\left\langle f^{*}, f^{*}\right\rangle_{\mathcal{K}}\right) \neq 0$ and $x \in X$ is arbitrary, this equality can be rewritten as

$$
f^{*}=-\frac{1}{2 g^{\prime}\left(\left\langle f^{*}, f^{*}\right\rangle_{\mathcal{K}}\right)} \sum_{i=1}^{N}\left(\partial_{i} \widetilde{L}\left(\mathbf{f}^{*}(\mathbf{x})\right)+\lambda^{\top} \partial_{i} \widetilde{I}\left(\mathbf{f}^{*}(\mathbf{x})\right)+\mu^{\top} \partial_{i} \widetilde{E}\left(\mathbf{f}^{*}(\mathbf{x})\right)\right) k\left(x_{i}, \cdot\right),
$$

which proves the statement (6). In particular, when $g(x)=c x$ for all $x \in \mathbb{R}$ we get

$$
f^{*}=-\frac{1}{2 c} \sum_{i=1}^{N}\left(\partial_{i} \widetilde{L}\left(\mathbf{f}^{*}(\mathbf{x})\right)+\lambda^{\top} \partial_{i} \widetilde{I}\left(\mathbf{f}^{*}(\mathbf{x})\right)+\mu^{\top} \partial_{i} \widetilde{E}\left(\mathbf{f}^{*}(\mathbf{x})\right)\right) k\left(x_{i}, \cdot\right)
$$

which shows that, in this case, there is no dependence on the choice of indefinite inner product $\langle\cdot, \cdot\rangle_{\mathcal{K}}$ associated to $k$.

We emphasise that, in view of this result, the situation for RKKS is analogous to the RKHS case in the sense that the solution of the stabilization problem is in the span of the data, although this adapted representer theorem is designed for the study of critical points, as opposed to minima of a loss functional, and thus algorithms must be adapted to this setting (Hassibi et al. (1999)). It turns out that they often can be.

Crucially, we have shown in Theorem 11, that in the case were $g$ is multiplication by a constant, one does not need access to the feature map $\Phi$, nor the RKKS $\mathcal{K}$ and its indefinite inner product $\langle\cdot, \cdot\rangle_{\mathcal{K}}$. So, in this case, one does not need access to the PD decomposition of $k$ to apply Theorem 11; one only needs to know of its existence.

The reason for stabilizing, as opposed to minimizing, the regularized loss functional $L(f, D)+g\left(\langle f, f\rangle_{\mathcal{K}}\right)$ is not that the proof of the RKKS representer theorem fails for minimization; it doesn't. Rather the reason is that it is not expected that the regularized loss functional admits a minimum, since $\langle f, f\rangle_{\mathcal{K}}$, as opposed to $\|f\|_{\mathcal{H}}$ in Theorem 4 , is not bounded below for $f \in \mathcal{K}$. Thus, solutions to learning problems of interest in RKKS will often take the form of a saddle point instead.

Example 12 Support vector machines (see Example 5) have been generalized to PD decomposable kernels in Loosli et al. (2016). Given a set $X$, a PD decomposable kernel $k$ on $X$, $\mathcal{K}$ an RKKS associated to $k$, a finite data set $D=\left\{\left(x_{1}, y_{1}\right), \ldots,\left(x_{N}, y_{N}\right)\right\} \subset(X \times\{-1,1\})^{N}$, the analogous version to the learning problem (3) can be written as

$$
\underset{f \in \mathcal{K}, b \in \mathbb{R}}{\substack{\text { stabilize }}}\langle f, f\rangle_{\mathcal{K}} \quad \text { with respect to } \quad \sum_{i=1}^{N} \max \left(0,1-y_{i}\left(f\left(x_{i}\right)+b\right)\right) \leq \tau
$$

for some fixed parameter $\tau>0$. The representer theorem tells us that any solution $(f, b)$ to this problem has the form $f=\sum_{i=1}^{N} \alpha_{i} k\left(x_{i}, \cdot\right)$ for some $\alpha_{1}, \ldots, \alpha_{N} \in \mathbb{R}$. We emphasise again that we are applying here the semiparametric extension to the representer theorem (Ong et al., 2004, Theorem 12), since the stabilization is over $f$ and $b$, and not solely over $f$ as in Theorem 11. See Figure 1 for an application.

Without a PD decomposition, we do not have a representer theorem like Theorem 11. In that case, it is not so much that the proof of the representer theorem fails, but rather that we cannot state it in the first place. Indeed, the kernel does not have a reproducing property as in Theorem 9, so there is no obvious concept of a space of functions we can stabilize over. Therefore, a solution found, for instance in Example 12, may not come with a guarantee to provide a good classification rule for unseen data. This gives rise to the following fundamental question:

Problem 13 When does a kernel $k$ admit a PD decomposition?
Remark 14 The existence of a PD decomposition guarantees that we can apply Theorem 11 for some RKKS $\mathcal{K}$ associated to $k$. As observed in Remark 10, a PD decomposable kernel may sometimes give rise to multiple distinct RKKS. Therefore, when such choice is not unique, the solution of the learning problem (5) depends on it. However, when $g$ is


Figure 1: The Krein SVM algorithm (Loosli et al. (2016)) applied on the hyperbolic plane $\mathbb{H}^{2}$, with the geodesic Gaussian kernel $k=\exp \left(-\lambda d(\cdot, \cdot)^{2}\right)$. The data is sampled from a Riemannian Gaussian distribution (Said et al. (2018, 2022, 2023)) centered at the origin of the Poincare disc, and is split into two classes according to geodesic decision boundaries (doted curves in the figure). The number of sampled data points is $200,200,500$ respectively. The results of the classification are displayed in the Poincare disc model of $\mathbb{H}^{2}$. We will show in Corollary 59 that the Gaussian kernel admits a PD decomposition on $\mathbb{H}^{2}$, justifying its use in this scenario.
multiplication by a constant, which is often the case in practice, we saw in Theorem 11 that the solution is not affected by this choice.

Problem 13 was first studied by Schwartz (1964). His setting was more general than ours, and only abstract conditions for PD decomposability were obtained. For instance, it is stated in that reference that $k$ has a PD decomposition if and only if there is a PD kernel $k_{+}$such that $k_{+}-k$ is PD (see (Schwartz, 1964, Proposition 38 and Proposition 23)). However, if one wants to recognise PD decomposable kernels in practice, this condition does not seem any easier to work with than Definition 7 itself.
Example 15 If $k_{1}, \ldots, k_{n}$ are PD kernels and $a_{1}, \ldots, a_{n} \in \mathbb{R}$, then the kernel

$$
\sum_{i=1}^{n} a_{i} k_{i}
$$

has the PD decomposition

$$
\sum_{i: a_{i} \geq 0} a_{i} k_{i}-\sum_{i: a_{i}<0}\left(-a_{i}\right) k_{i} .
$$

In fact, observe that the space of kernels that admit a PD decomposition is exactly the real span of the PD kernels.

Example 16 The pointwise limit of PD kernels is PD. It follows that the pointwise limit of PD decomposable kernels is PD decomposable.

Remark 17 PD kernels form a convex cone that is closed under pointwise convergence. Now, Example 15 and Example 16 show that PD decomposable kernels are the real linear span of these, and thus form a real vector space closed under pointwise convergence.

Example 18 Similarly, since the pointwise product of PD kernels is PD (Berg et al., 1984, Chapter 3 Theorem 1.12), so it follows that the pointwise product of PD decomposable kernels is PD decomposable.

Example 19 Suppose $X$ is finite. Then a kernel $k$ can be viewed as a Hermitian matrix $K$. By Sylvester's law of inertia, we can write

$$
K=A^{\dagger} E A
$$

for some invertible real matrix $A$, where $A^{\dagger}$ denotes the conjugate transpose of $A$, and $E$ is of the form

$$
E=\left[\begin{array}{lll}
I_{N_{+}} & & \\
& -I_{N_{-}} & \\
& & 0_{M}
\end{array}\right]
$$

for some $N_{+}, N_{-}, M \in \mathbb{Z}_{\geq 0}$, where $I_{N+}, I_{N-}$ denote the identity matrices with the corresponding sizes, and $0_{M}$ is the zero matrix of size $M \times M$. Then $E=E_{+}-E_{-}$where

$$
E_{+}=\left[\begin{array}{lll}
0_{N_{+}} & & \\
& -I_{N_{-}} & \\
& & 0_{M}
\end{array}\right], E_{-}=\left[\begin{array}{lll}
I_{N_{+}} & & \\
& 0_{N_{-}} & \\
& & 0_{M}
\end{array}\right] .
$$

This implies that

$$
K=A^{\dagger} E_{+} A-A^{\dagger} E_{-} A,
$$

and hence $k$ has a PD decomposition. However, when $X$ is infinite, one should not expect all kernels to admit a PD decomposition.

Example 20 We present an example of a finite-dimensional RKKS associated with a nonfinite $X$, namely $X=\mathbb{H}^{n}$, the $n$-dimensional hyperbolic space. In order to motivate our construction, we recall that when $X=\mathbb{R}^{n}$, the standard inner product kernel $\langle\cdot, \cdot\rangle: \mathbb{R}^{n} \times$ $\mathbb{R}^{n} \rightarrow \mathbb{R}$ is PD and gives rise to the RKHS $\mathcal{H}=\left\{\langle x, \cdot\rangle: x \in \mathbb{R}^{n}\right\}$, which consists of all linear maps $\mathbb{R}^{n} \rightarrow \mathbb{R}$. So for $f \in \mathcal{H}, b \in \mathbb{R}$, the set $\left\{x \in \mathbb{R}^{n}: f(x)=b\right\}$ is a hyperplane in $\mathbb{R}^{n}$.

Taking $X=\mathbb{H}^{n}$, it is therefore natural to ask if there is a kernel that gives rise to an RKHS of functions $f$ such that for any $b \in \mathbb{R}$, the sets $\left\{x \in \mathbb{H}^{n}: f(x)=b\right\}$ are geodesic hyperplanes. This has practical relevance, as it would allow us to perform large-margin classification with geodesic decision boundaries, as in Cho et al. (2019). In fact this is done in a natural way with a PD decomposable kernel giving rise to an RKKS. First consider the Minkowski inner product on $\mathbb{R}^{n+1}$ :

$$
(x, y)=x_{0} y_{0}-x_{1} y_{1}-\cdots-x_{n} y_{n}
$$

for all $x=\left(x_{0}, \ldots, x_{n}\right), y=\left(y_{0}, \ldots, y_{n}\right) \in \mathbb{R}^{n+1}$. Then, we can view $\mathbb{H}^{n}$ as a hypersurface in this Minkowski space:

$$
\mathbb{H}^{n}=\left\{x \in \mathbb{R}^{n+1}:(x, x)=1, x_{0}>0\right\} .
$$

This is the hyperboloid model of $\mathbb{H}^{n}$. The geodesic hyperplanes in $\mathbb{H}^{n}$ are given by

$$
\left\{x \in \mathbb{H}^{n}:(y, x)=b\right\}
$$



Figure 2: A geodesic in the hyperboloid model of the hyperbolic plane $\mathbb{H}^{2}$, obtained as the intersection of a plane with the hyperboloid. The corresponding geodesic on the Poincare disc model $\mathbb{B}^{2}$ of the hyperbolic plane is obtained by stereographic projection from the point $(0,0,-1)$.
for some $y \in \mathbb{R}^{n+1}$, where the above is non-empty if and only if $(y, y)<0$. These are the intersection of planes through the origin with the hyperboloid model of $\mathbb{H}^{n}$ (see Figure 2). So the kernel $k: \mathbb{H}^{n} \times \mathbb{H}^{n} \rightarrow \mathbb{R}$ of interest can be taken to be $k=(\cdot, \cdot)$. It has a PD decomposition by definition of the Minkowski inner product, and the associated real RKKS is the Minkowski space $\mathbb{R}^{n+1}$ equipped with $(\cdot, \cdot)$.

Example 21 We now present a kernel that does not admit a PD decomposition. Let $B$ be a real reflexive Banach space whose norm does not arise from an inner product (for instance an $L^{p}$ space with $\left.1<p<\infty, p \neq 2\right), B^{*}$ be its dual. Let $X=B \times B^{*}$, and let the kernel $k: X \times X \rightarrow \mathbb{R}$ be given by

$$
k(x, \varphi ; y, \psi):=\varphi(y)+\psi(x) .
$$

Then $k$ has no PD decomposition. This example was first presented in (Schwartz, 1964, Page 243). A self-contained proof can be found in (Alpay, 1991, Theorem 2.2).

## 3 Group Actions and Invariant Kernels

In order to have at our disposal additional structure to address Problem 13, we shall restrict our attention to kernels that possess certain symmetries, which we shall refer to as invariant kernels. In the rest of the paper, $G$ will be a locally compact (always assumed Hausdorff) topological group, acting on the left on the set $X$.

Definition $22 A$ kernel $k: X \times X \rightarrow \mathbb{C}$ is called invariant with respect to the action of $G$ if

$$
k(g \cdot x, g \cdot y)=k(x, y)
$$

for all $g \in G$ and all $x, y \in X$, where the dot $\cdot$ denotes a left action of the elements of $G$ on the elements of $X$.

PD invariant kernels were first studied by Yaglom (1961), and more recently by Azangulov et al. (2023b a) in the context of Gaussian processes.

Example 23 Taking $X=G=\mathbb{R}^{n}$ equipped with addition and acting on itself, an invariant kernel on $\mathbb{R}^{n}$ is a translation-invariant or stationary kernel:

$$
k(x+g, y+g)=k(x, y)
$$

for all $x, y, g \in \mathbb{R}^{n}$. Standard invariant theory arguments show that translation-invariant kernels are exactly those that that depend only on the difference $x-y$, i.e. kernels $k$ on $\mathbb{R}^{n}$ for which there is a function $f$ on $\mathbb{R}^{n}$ such that

$$
k(x, y)=f(x-y)
$$

for all $x, y \in \mathbb{R}^{n}$.
In this paper, we shall study a generalization of this example, namely when the action of $G$ is transitive on $X$. In that case, if we pick a distinguished element $o$ of $X$, the orbit-stabilizer theorem gives us the bijection of sets

$$
\begin{align*}
G / H & \rightarrow X \\
g H & \rightarrow g \cdot o . \tag{8}
\end{align*}
$$

where $H=\operatorname{Stab}(o)$. In what follows, we will write $X$ as $G / H$, where $H$ is a subgroup of $G$. This notation has the advantage of specifying both the group $G$ acting on $X$ and the action (the canonical left action of $G$ on $G / H)$, so from here on we will call an invariant kernel on $G / H$ one that is invariant with respect to that action. We will further make the assumption that $H$ is closed, granting us with a locally compact Hausdorff quotient topology for $X=G / H$, and making the action of $G$ on $X$ proper (see (Folland, 2015, Section 2.6) for the theory of such homogeneous spaces).

Invariant kernels under transitive actions are particularly interesting for us because the symmetries they respect allow us to view them as functions of a single variable instead of two, as in Example 23. To see this, let us first give the analogues to Definitions 1 and 7 for functions defined on the relevant spaces.

Definition 24 Write $H \backslash G / H:=\{H g H: g \in G\}$ for the double coset space. A complexvalued function $f: H \backslash G / H \rightarrow \mathbb{C}$ is said to be positive-definite (PD) if for all $N \in \mathbb{N}$ and $g_{1}, \ldots, g_{N} \in G$, the matrix

$$
\left(f\left(H g_{i}^{-1} g_{j} H\right)\right)_{i, j}
$$

is positive semi-definite. $f$ is said to have a positive (PD) decomposition if it can be written as

$$
f=f_{+}-f_{-}
$$

where $f_{+}$and $f_{-}$are $P D$ functions on $H \backslash G / H$.


Figure 3: A visualisation of the projections to the double coset spaces $\pi$ : $S^{2} \cong$ $S O(3) / S O(2) \rightarrow S O(2) \backslash S O(3) / S O(2)$ and $\pi: \mathbb{S}_{++}^{2} \cong G L(2) / O(2) \rightarrow$ $O(2) \backslash G L(2) / O(2)$. In each case, we have in red an orbit $H g H$. In the case of $S^{2}$, it is obtained by rotating $g S O(2)$ around a horizontal axis. In the case of $\mathbb{S}_{++}^{2}$, it is the set of matrices with eigenvalues equal to the ones of $g O(2)$, obtained as the intersection of the constant-trace plane and the constant-determinant surface. In each case, we have in dark blue a set transversal to such orbits, and homeomorphic to $H \backslash G / H$.

In particular, when $H=\{e\}, f: G \rightarrow \mathbb{C}$ is PD if the matrices $\left(f\left(g_{i}^{-1} g_{j}\right)\right)_{i, j}$ are positive semi-definite for all $N$ and all $g_{i}, g_{j} \in G$. Functions on $H \backslash G / H$ should be considered as $H$-invariant functions on $G / H$. The relevance of double coset spaces in the kernel context will become clear later in Proposition 29.

Example 25 The $n$-dimensional sphere $S^{n}$ may be viewed as the coset space

$$
S^{n} \cong S O(n+1) / S O(n)
$$

Moreover, viewed as a submanifold $S^{n} \subset \mathbb{R}^{n+1}, S O(n+1)$ acts transitively on $S^{n}$ by rotations. Picking $o=(1,0, \ldots, 0) \in S^{n}$, the stabilizer of $o$ is the subgroup of $S O(n+1)$ leaving the first coordinate untouched, which is isomorphic to $S O(n)$. Now the double coset space $S O(n) \backslash S O(n+1) / S O(n)$ is obtained by quotienting out each orbit (see Figure 3), so we see

$$
S O(n) \backslash S O(n+1) / S O(n) \cong[-\pi, \pi] .
$$

Functions on $S O(n) \backslash S O(n+1) / S O(n)$ are functions of the geodesic distance on $S^{n}$, or equivalently of the inner product of $\mathbb{R}^{n+1}$ restricted to $S^{n}$.

Example 26 The space of $n \times n$ real symmetric positive-definite matrices $\mathbb{S}_{++}^{n}$ may be viewed as the coset space

$$
\mathbb{S}_{++}^{n} \cong G L(n) / O(n)
$$

Indeed, $G L(n)$ acts transitively on $\mathbb{S}_{++}^{n}$ by $G \cdot X:=G X G^{\top}$ for $G \in G L(n)$ and $X \in$ $\mathbb{S}_{++}^{n}$. The stabilizer of the the identity $I$ consists of the matrices $G$ such that $G G^{\top}=I$,
i.e. $\operatorname{Stab}(I)=O(n)$. Now, for $X \in \mathbb{S}_{++}^{n}, O(n) X O(n) \in O(n) \backslash G L(n) / O(n)$ consists of $\left\{H X H^{\top}: H \in O(n)\right\}$, which is fully determined by the $n$ eigenvalues of $X$. So

$$
O(n) \backslash G L(n) / O(n) \cong\left\{\left(\lambda_{1}, \ldots, \lambda_{n}\right): \lambda_{i}>0 \forall i\right\} / \sim
$$

where $\left(\lambda_{1}, \ldots, \lambda_{n}\right) \sim\left(\lambda_{1}^{\prime}, \ldots, \lambda_{n}^{\prime}\right)$ if and only if $\left\{\lambda_{1}, \ldots, \lambda_{n}\right\}=\left\{\lambda_{1}^{\prime}, \ldots, \lambda_{n}^{\prime}\right\}$ (see Figure 3 ). Functions on $O(n) \backslash G L(n) / O(n)$ are functions of the eigenvalues on $\mathbb{S}_{++}^{n}$.

Let us generalize the concept of Hermitian functions to complex-valued functions on double coset spaces:

Definition $27 f: H \backslash G / H \rightarrow \mathbb{C}$ is called Hermitian if

$$
f\left(H g^{-1} H\right)=\overline{f(H g H)}
$$

for all $g \in G$.
Since $G$ is a topological group, $G / H$ and $H \backslash G / H$ are equipped with natural quotient topologies. In the rest of the paper, we shall make the following continuity assumption: we will consider continuous kernels $k$, and ask whether they admit PD decompositions into continuous PD kernels $k=k_{+}-k_{-}$. We call such a decomposition a continuous $P D$ decomposition.

Remark 28 The continuity assumption is non-trivial: see (Schaback and Wendland, 2003, Section 3) for a construction of an everywhere discontinuous PD kernel. Note, however, that at least for functions on groups, a continuous function on $G$ has a continuous PD decomposition if and only if it has a PD decomposition (Kaniuth and Lau, 2018, Corollary 2.2.2).

The reason for considering functions on the double coset space $H \backslash G / H$ as opposed to the regular coset space $G / H$ is made apparent in the following important result.

Proposition 29 There is a bijection between the continuous Hermitian functions on $H \backslash G / H$ and the continuous invariant kernels on $G / H$. This bijection restricts to:

1. A bijection between the continuous Hermitian $P D$ functions and the continuous $P D$ invariant kernels,
2. An injection between the continuous Hermitian $P D$ functions with continuous $P D$ decompositions and the continuous kernels with continuous PD decompositions.

Proof Given a continuous invariant kernel $k: G / H \times G / H \rightarrow \mathbb{C}$, define the function $f: H \backslash G / H \rightarrow \mathbb{C}$ by

$$
f(H g H):=k(g H, e H)
$$

for all $g \in G$, where $e$ is the identity of $G$. Note that

$$
f\left(H g^{-1} H\right)=k\left(g^{-1} H, e\right)=k\left(g^{-1} H, e H\right)=k(e H, g H)=\overline{k(g H, e H)}=\overline{f(H g H)}
$$

i.e. if $k$ is Hermitian then $f$ is necessarily Hermitian. Finally, the continuity of $f$ follows from the continuity of $k$. Conversely, given a Hermitian function $f: H \backslash G / H \rightarrow \mathbb{C}$, define the invariant kernel $k: G / H \times G / H \rightarrow \mathbb{C}$ by

$$
k\left(g H, g^{\prime} H\right):=f\left(H g^{-1} g^{\prime} H\right)
$$

for all $g, g^{\prime} \in G$ where the continuity of and the Hermitian character of $f$ imply the same properties for $k$.

These correspondences are inverses of each other and thus provide a bijection. Moreover, we can see that $k$ is PD if and only if $f$ is PD . Also, this correspondence is linear, which implies that if $f$ admits a continuous PD decomposition, so does $k$.

Remark 30 In the case $H=\{e\}$, the correspondence above is simply the bijection between Hermitian functions on $G$ and invariant kernels on $G$. Example 23 is an instance of this case.

Remark 31 Note that in the second part of Proposition 29, we have only proved the injectivity of the correspondence between continuous Hermitian PD functions with continuous PD decompositions and the continuous kernels with continuous PD decompositions. The difficulty in proving surjectivity is that $k$ may not admit a continuous PD decomposition $k=k_{+}-k_{-}$with $k_{+}$and $k_{-}$invariant. Later on in Section 5.1, we shall face a similar problem for which we provide a solution. Additionally, in Appendix A we will show that the correspondence in part 2 is indeed surjective when $G$ is compact. In any case, the injectivity in this part will be enough for us to provide sufficient conditions for a kernel $k$ to have a continuous PD decomposition.

Being able to handle invariant kernels as functions will allow us to leverage tools from harmonic analysis and representation theory to address Problem 13.

In the sequel, we shall use the following spaces in relation with the locally compact group $G$ and the closed subgroup $H$ :

$$
\begin{aligned}
B_{+}(G) & :=\{\text { continuous PD functions } G \rightarrow \mathbb{C}\} \\
B_{\mathbb{R}}(G) & :=B_{+}(G) \otimes \mathbb{R} \\
& =\{\text { continuous functions with continuous PD decompositions } G \rightarrow \mathbb{C}\} \\
B_{\mathbb{C}}(G) & :=B_{+}(G) \otimes \mathbb{C}
\end{aligned}
$$

and, more generally,
$B_{+}(H \backslash G / H):=\{\mathrm{PD}$ functions $H \backslash G / H \rightarrow \mathbb{C}\}$
$B_{\mathbb{R}}(H \backslash G / H):=B_{+}(H \backslash G / H) \otimes \mathbb{R}$
$=\{$ continuous functions with continuous PD decompositions $H \backslash G / H \rightarrow \mathbb{C}\}$
$B_{\mathbb{C}}(H \backslash G / H):=B_{+}(H \backslash G / H) \otimes \mathbb{C}$.

Remark 32 While $B_{\mathbb{R}}(G)$ and $B_{\mathbb{R}}(H \backslash G / H)$ are defined as the real span of $B_{+}(G)$ and $B_{+}(H \backslash G / H)$, note that the latter contain complex-valued functions that are not real-valued, hence so do the former.

The space $B_{\mathbb{C}}(G)$ is called the Fourier-Stieltjes algebra of $G$ in the literature and has been studied thoroughly; see, for example, Kaniuth and Lau (2018). We will present and extend some of these results to $B_{\mathbb{C}}(H \backslash G / H)$.

Motivated by Proposition 29, our goal is to characterize $B_{\mathbb{R}}(H \backslash G / H)$ and consider Hermitian elements in this space. In fact, since we are interested in real-valued kernels and hence in real-valued functions, the following proposition shows that it suffices to characterize $B_{\mathbb{C}}(H \backslash G / H)$ and consider real-valued even $\left(f(H g H)=f\left(H g^{-1} H\right)\right.$ for all $\left.g\right)$ elements in this space.

Proposition 33 If $f \in B_{\mathbb{C}}(G)$ or $f \in B_{\mathbb{C}}(H \backslash G / H)$ is real-valued, then $f \in B_{\mathbb{R}}(G)$ or $f \in B_{\mathbb{R}}(H \backslash G / H)$ respectively.

Proof Observe that if $f \in B_{+}(G)$, then $\operatorname{Re} f=\frac{f+\bar{f}}{2} \in B_{+}(G)$, and $\operatorname{Im} f=\frac{f-\bar{f}}{2} \in B_{\mathbb{R}}(G)$. If $f \in B_{\mathbb{C}}(G)$ we can write

$$
f=f_{+}-f_{-}+i f_{+i}-i f_{-i}
$$

with $f_{+}, f_{-}, f_{+i}, f_{-i} \in B_{+}(G)$. If $f$ is real-valued,

$$
f=\operatorname{Re} f=\operatorname{Re} f_{+}-\operatorname{Re} f_{-}+\operatorname{Re}\left(i f_{+i}\right)-\operatorname{Re}\left(i f_{-i}\right)=\operatorname{Re} f_{+}-\operatorname{Re} f_{-}-\operatorname{Im} f_{+i}+\operatorname{Im} f_{-i}
$$

so $f \in B_{\mathbb{R}}(G)$. The proof is identical for functions in $B_{\mathbb{C}}(H \backslash G / H)$.

## 4 Positive Decompositions for Invariant Kernels: Commutative Case

In this section, we focus on invariant kernels that are invariant with respect to the action of an Abelian locally compact group $G$. The commutativity hypothesis will simplify the theory considerably.

First, observe that in this case, any closed subgroup $H$ of $G$ is automatically a normal subgroup, so $X=G / H$ is itself a group. Equipped with the quotient topology, $G / H$ is actually a locally compact group (Folland, 2015, Proposition 2.2). Therefore without loss of generality, we can set $X=G$.

In view of the results in the previous section, we shall focus in this case on $B_{\mathbb{C}}(G)$, which is a well-studied object in standard commutative abstract harmonic analysis. For the sake of completeness, we recall here a few facts in relation to this theory that are needed in our developments. We refer to Folland (2015) for a more detailed exposition and proofs. Chapter 4 in Folland (2015) is particularly relevant to this section.
$G$ has a dual group, $\mathcal{G}$, which is defined as the group of continuous irreducible unitary representations of $G$. $G$ being Abelian implies that irreducible unitary representations of $G$ are one-dimensional, so $\widehat{G}$ consists of the continuous group homomorphisms $\xi: G \rightarrow S^{1} \subset \mathbb{C}$. Equipped with the open-compact topology, $\widehat{G}$ is itself a locally compact group. Moreover,
we have a natural topological group isomorphism $\widehat{\widehat{G}} \cong G$. This is the Pontryagin Duality Theorem.

Now define

$$
\begin{aligned}
M_{+}(G) & :=\{\text { finite positive Radon measures on } G\} \\
M_{\mathbb{R}}(G) & :=M_{+}(G) \otimes \mathbb{R} \\
& =\{\text { finite signed Radon measures on } G\} \\
M_{\mathbb{C}}(G) & :=M_{+}(G) \otimes \mathbb{C} \\
& =\{\text { finite complex Radon measures on } G\}
\end{aligned}
$$

and $M_{+}(\widehat{G}), M_{\mathbb{R}}(\widehat{G}), M_{\mathbb{C}}(\widehat{G})$ analogously. Importantly for us, we can define the Fourier transform $\mathscr{F}: M_{\mathbb{C}}(G) \rightarrow C(\widehat{G})$ by

$$
\mathscr{F} \mu(\xi)=\int_{G} \overline{\xi(g)} d \mu(g)
$$

for all $\xi \in \widehat{G}$. Here $C(\widehat{G})$ is the space of continuous complex-valued functions on $\widehat{G}$. Similarly, we can define the inverse Fourier transform $\mathscr{F}^{-1}: M_{\mathbb{C}}(\widehat{G}) \rightarrow C(G)$ by

$$
\mathscr{F}^{-1} \nu(g)=\int_{G} \xi(g) d \nu(\xi)
$$

for all $g \in G$. $\mathscr{F}$ and $\mathscr{F}^{-1}$ are injective linear maps.
Fourier transforms are often thought of as acting on functions. Our description of Fourier transforms encompasses this: note that $G$ comes equipped with a Haar measure $\mu$. This is a translation-invariant Radon measure on $G$, which is unique up to multiplication by a positive constant. Then, for $f$ a $\mu$-integrable function on $G, f \mu \in M_{\mathbb{C}}(G)$, so we can define the Fourier transform of $f$ as $\mathscr{F}(f \mu)$.

Remark $34 \mathscr{F}$ and $\mathscr{F}^{-1}$ will only be inverse to each other with the right choice of Haar measures on $G$ and $\widehat{G}$. Indeed, as described above, the actions of the operators $\mathscr{F}, \mathscr{F}^{-1}$ on $L^{1}(G), L^{1}(\widehat{G})$ depend on the choice of Haar measures. See (Folland, 2015, Theorem 4.22).

Example 35 When $G=\mathbb{R}^{n}$, we have $\widehat{G} \cong \mathbb{R}^{n}$, and the Fourier transform can be written as

$$
\mathscr{F} \mu(\xi)=\int_{\mathbb{R}^{n}} e^{-i\langle\xi, x\rangle} d \mu(x)
$$

for $\mu \in M_{\mathbb{C}}\left(\mathbb{R}^{n}\right)$ and $\xi \in \mathbb{R}^{n}$. The inverse Fourier transform can be written as

$$
\mathscr{F}^{-1} \nu(x)=\int_{\mathbb{R}^{n}} e^{i\langle\zeta, x\rangle} d \nu(\xi)
$$

for $\nu \in M_{\mathbb{C}}\left(\mathbb{R}^{n}\right)$ and $x \in \mathbb{R}^{n}$.
Example 36 When $G=S^{1}$, we have $\widehat{G} \cong \mathbb{Z}$, and the Fourier transform corresponds to the discrete Fourier transform, and can be written as

$$
\mathscr{F} \mu(k)=\int_{S^{1}} e^{-i k[x]} d \mu(x)
$$

for $\mu \in M_{\mathbb{C}}\left(S^{1}\right)$ and $k \in \mathbb{Z}$, where $[x] \in[-\pi, \pi)$ is the argument of $x$. The inverse Fourier transform can be written as

$$
\mathscr{F} \nu(x)=\int_{S^{1}} e^{i k[x]} d \nu(x)
$$

for $\nu \in M_{\mathbb{C}}\left(S^{1}\right)$ and $x \in S^{1}$.
We are now able to state the key theorem of this section (Folland, 2015, Theorem 4.19).
Theorem 37 (Bochner's Theorem) $f \in B_{+}(G)$ is PD if and only if there is $\nu \in M_{+}(\widehat{G})$ such that $f=\mathscr{F}^{-1} \nu$. Moreover, such $\nu$ is unique.

In other words, $\mathscr{F}^{-1}$ is a linear isomorphism between $M_{+}(\widehat{G})$ and $B_{+}(G)$ (which are convex cones, not vector spaces). Therefore, tensoring by $\mathbb{R}$ and then by $\mathbb{C}$ in Bochner's Theorem, we also get that $\mathscr{F}^{-1}$ is a linear isomorphism between $M_{\mathbb{R}}(\widehat{G}), M_{\mathbb{C}}(\widehat{G})$ and $B_{\mathbb{R}}(G), B_{\mathbb{C}}(G)$, respectively. In other words, we have the commutative diagram

where $\sim$ denote linear isomorphisms and $\hookrightarrow$ denote the inclusions. In other words, PD decomposable functions on $G$ are the functions that are the inverse Fourier transforms of finite signed measures on $\widehat{G}$. This description of $B_{\mathbb{R}}(G)$ and $B_{\mathbb{C}}(G)$ is in some sense the best we have, as being the inverse Fourier transform of a finite measure does not appear to be a reducible property. However, what we can do is provide sufficient conditions for a function to be in $B_{\mathbb{R}}(G)$ and $B_{\mathbb{C}}(G)$.

First, let us state two important results.
Theorem 38 The map $\mathscr{F}^{-1}: M(\widehat{G}) \rightarrow C(G)$ is surjective if and only if $G$ is finite. So, $B_{\mathbb{C}}(G)=C(G)$ if and only if $G$ is finite.

See Graham and McGehee (1979). We showed the if direction in Example 19, and this proposition provides a converse. It also provides a glimpse at the fact that characterizing $B_{\mathbb{C}}(G)$ is a deep problem.

We have, however, the following positive result:
Theorem $39 B_{\mathbb{C}}(G) \cap C_{c}(G)$ is dense in $C_{c}(G)$, the space of compactly supported continuous functions, with the uniform norm. As a consequence, $B_{\mathbb{C}}(G) \cap L^{p}(G)$ is dense in $L^{p}(G)$ for all $1 \leq p<\infty$.

See (Folland, 2015, Proposition 3.33). Theorem 45 below is the general version of this result for the non-commutative case.

For the commutative case, there are, in practice, two examples of interest: $\mathbb{R}$ and $S^{1}$. This is reflected by the fact that any Abelian Lie group $G$ is a product of these. In Example 35 and Example 36, we have already given expressions for the Fourier transform in these cases. Let us study the implications in each case.
Example $40 G=\mathbb{R}^{n}, \widehat{G}=\mathbb{R}^{n}$. In this case, it is well-known that $\mathscr{F}$ is an automorphism of the Schwartz space $\mathscr{S}\left(\mathbb{R}^{n}\right)$. So, in particular, $\mathscr{S}\left(\mathbb{R}^{n}\right) \subset B_{\mathbb{C}}\left(\mathbb{R}^{n}\right)$, where we recall that

$$
\mathscr{S}\left(\mathbb{R}^{n}\right):=\left\{f \in C^{\infty}\left(\mathbb{R}^{n}\right): \sup _{x \in \mathbb{R}^{n}}\left|x^{\alpha}\left(D^{\beta} f\right)(x)\right|<\infty \forall \alpha, \beta \in \mathbb{N}^{n}\right\}
$$

In other words, $\mathscr{S}\left(\mathbb{R}^{n}\right)$ is the space of complex-valued smooth functions on $\mathbb{R}^{n}$ whose values and derivatives decay faster at infinity than any reciprocal of a polynomial.

Example 41 We can view $S^{1}$ as $G=\mathbb{R} / \mathbb{Z}$, and we have $\widehat{G}=\mathbb{Z}$. As described in Example 36 , the Fourier transform in this case is the discrete Fourier transform. We have that $M_{\mathbb{C}}(\mathbb{Z})$ is $l_{\mathbb{C}}^{1}(\mathbb{Z})$, and $B_{\mathbb{C}}(\mathbb{R} / \mathbb{Z})$ is the space of absolutely continuous Fourier series. This is called the Wiener algebra, and we have that the Hölder continuous functions $C^{\alpha}(\mathbb{R} / \mathbb{Z}) \subset B_{\mathbb{C}}(\mathbb{R} / \mathbb{Z})$ for $\alpha>1 / 2$. This is Bernstein's Theorem, see for example (Katznelson, 2004, Theorem 6.3). Recall that for $\alpha<1$,

$$
C^{\alpha}(\mathbb{R} / \mathbb{Z}):=\left\{f \in C(\mathbb{R} / \mathbb{Z}): \sup _{x, y \in \mathbb{R} / \mathbb{Z}, x \neq y} \frac{|f(x)-f(y)|}{|x-y|^{\alpha}}<\infty\right\}
$$

where the absolute value $|x-y|$ in the denominator is taken modulo 1 .
Example 42 The Gaussian kernel on the circle $S^{1}$

$$
k(x, y)=\exp \left(-\lambda|x-y|^{2}\right)
$$

for all $x, y \in \mathbb{R} / \mathbb{Z} \cong S^{1}$, where $\lambda>0$ and the absolute value is taken modulo 1 , is not PD for any $\lambda>0$ (see Da Costa et al. (2023)). However, note that it is invariant, and under Proposition 29 corresponds to the Gaussian function $f(x)=\exp \left(-\lambda|x|^{2}\right)$ for all $x \in \mathbb{R} / \mathbb{Z} \cong S^{1}$, where the absolute value is taken modulo $1 . f$ is Lipschitz, so by Example $41, f \in B_{\mathbb{C}}\left(S^{1}\right)$, and since it is real-valued, $f \in B_{\mathbb{R}}\left(S^{1}\right)$ by Proposition 33. So $k$, despite not being PD, has a PD decomposition.

From Example 42, we obtain the following result.
Proposition 43 The Gaussian kernel $k=\exp \left(-\lambda d(\cdot, \cdot)^{2}\right)$ has a PD decomposition on any Abelian Lie group (with appropriate choice of the metric d) for any $\lambda>0$.

Proof Any Abelian Lie group $G$ is a product of copies of $\mathbb{R}$ and $S^{1}$, say $G \cong \mathbb{R}^{n} \times$ $\left(S^{1}\right)^{m}$. Taking the product Riemannian metric on the right-hand side of this induced by the standard Riemannian metrics on $\mathbb{R}$ and $S^{1}$, we can write the Gaussian kernel $k$ on $G$ as

$$
k(x, y)=\exp \left(-\lambda d(x, y)^{2}\right)=\prod_{i=1}^{n+m} \exp \left(-\lambda\left|x_{i}-y_{i}\right|^{2}\right)
$$

for all $x, y \in G \cong \mathbb{R}^{n} \times\left(S^{1}\right)^{m}$, where the last $m$ absolute values are taken modulo 1 . By positive-definiteness of the Gaussian kernel on $\mathbb{R}$ and PD decomposability of the Gaussian kernel on $S^{1}$ (Example 42), $k$ is a product of PD decomposable kernels, so has a PD decomposition by Example 18.
We conclude this section with the following example.
Example 44 Suppose $\tilde{f} \in C^{\alpha}(\mathbb{R})$ for some $\alpha>1 / 2$ and that $\tilde{f}$ is periodic, with period $T$, say. Then, if $\pi: \mathbb{R} \rightarrow \mathbb{R} / T \mathbb{Z}$ is the projection, there is a function $f \in C^{\alpha}(\mathbb{R} / T \mathbb{Z})$ such that $f \circ \pi=\tilde{f}$. Then, $f \in B_{\mathbb{C}}(\mathbb{R} / T \mathbb{Z})$, assuming for simplicity that $\tilde{f}$ is real-valued. Writing its PD decomposition $f=f_{+}-f_{-}$, we have that $\tilde{f}=f_{+} \circ \pi-f_{-} \circ \pi$ is a PD decomposition, so that $\tilde{f} \in B_{R}(\mathbb{R})$.

PD decompositions behave well with respect to quotients. We will prove a general version of this phenomenon, as well as its converse, in Section 5.1.

## 5 Positive Decompositions for Invariant Kernels: Non-Commutative Case

Armed with an understanding of the commutative case, we are now in a position to tackle the non-commutative case. Recall that we aim to characterize sufficient conditions for an invariant kernel to admit a PD decomposition, which by Proposition 29 and Proposition 33, comes down to characterizing sufficient conditions for a function to belong to $B_{\mathbb{C}}(H \backslash G / H)$. Here the only assumptions are that $G$ is a locally compact group and $H$ is a closed subgroup of $G$.

We begin by observing that, just as in the commutative case (Theorem 39), PD decomposable functions on $G$ are plentiful and are, in fact, dense in the space of compactly supported continuous functions:

Theorem $45 B_{\mathbb{C}}(G) \cap C_{c}(G)$ is dense in $C_{c}(G)$, the space of compactly supported continuous functions, with the uniform norm. As a consequence, $B_{\mathbb{C}}(G) \cap L^{p}(G)$ is dense in $L^{p}(G)$ for all $1 \leq p<\infty$.

See (Kaniuth and Lau, 2018, Corollary 2.3.5) for a proof.
To obtain more specific descriptions of $B_{\mathbb{C}}(G)$ and $B_{\mathbb{C}}(H \backslash G / H)$, there are multiple different possible approaches. Firstly, Fourier transforms can be defined on general compact groups $G$ analogously to the circle: they can be described as Fourier series, see (Folland, 2015, Chapter 5). The non-compact case is more difficult. Additional regularity assumptions must be placed on the group, for example, that it is of type I, see for instance Yaglom (1961). This latter reference provides Bochner-like theorems for the homogeneous spaces of type I groups, which include symmetric spaces. The most general version of Bochner's Theorem is known as the Bochner-Godement Theorem, see (Faraut and Harzallah, 1974, Theorem 3.1), which describes PD $H$-invariant functions on $G / H$, i.e. PD functions on $H \backslash G / H$, with some additional mild regularity assumptions on $G / H$.

All of these references are quite technical. In this paper we will take a different route. Describing a general Bochner Theorem for non-Abelian groups will not get us much closer to characterizing $B_{\mathbb{C}}(H \backslash G / H)$. What we are really after is concrete sufficient conditions for a function to belong to $B_{\mathbb{C}}(H \backslash G / H)$ on spaces of interest, for instance when $G$ is a Lie group and $H$ is a Lie subgroup. Therefore, we will proceed as follows.

As exemplified by Theorem $45, B_{\mathbb{C}}(G)$ has been more extensively studied than the more general $B_{\mathbb{C}}(H \backslash G / H)$. We will therefore start in Section 5.1 by describing how the latter relates to the former. After this we will be in a position in Section 5.2 to describe sufficient conditions for a function to belong to $B_{\mathbb{C}}(H \backslash G / H)$ for general classes of Lie groups $G$ and Lie subgroups $H$.

### 5.1 Positive decompositions through representation theory

Suppose that $f \in C(H \backslash G / H)$. If $f \in B_{\mathbb{R}}(H \backslash G / H)$ and $\pi: G \rightarrow H \backslash G / H$ is the projection, write

$$
f=f_{+}-f_{-}
$$

with $f_{+}, f_{-} \in B_{+}(H \backslash G / H)$. Then

$$
f \circ \pi=f_{+} \circ \pi-f_{-} \circ \pi
$$

with $f_{+} \circ \pi, f_{-} \circ \pi \in B_{+}(H \backslash G / H)$, so $f \circ \pi \in B_{\mathbb{R}}(G)$.
If we can characterize $B_{\mathbb{R}}(G)$ well, then it is the converse we would like to show: that if $f \in C(H \backslash G / H)$ and $f \circ \pi \in B_{\mathbb{R}}(G)$, then $f \in B_{\mathbb{R}}(H \backslash G / H)$. This is not obvious: from the fact that $f$ admits a PD decomposition with functions on $G$ and is invariant on double cosets, we would like to show that we can choose a PD decomposition for $f$ for which the terms are themselves invariant on double cosets.

One idea may be, given $f \in C(H \backslash G / H)$ with $f \circ \pi \in B_{\mathbb{R}}(G)$ and a PD decomposition $f \circ \pi=\tilde{f}_{+}-\tilde{f}_{-}$, to "average out" values of $\tilde{f}_{+}$and $\tilde{f}_{-}$along each double coset, in order to find $\mathrm{PD} f_{+}, f_{-} \in C(H \backslash G / H)$ such that $f=f_{+}-f_{-}$. This argument works if $H$ is compact, and we make this proof precise in Appendix A.

For the general case, the situation is more difficult, as we cannot simply average over the non-compact double cosets. It will be the goal of this section to provide an alternative general proof, and we will do so using the language of representation theory.

Definition 46 A representation of $G$ is a group homomorphism

$$
\pi: G \rightarrow G L_{\mathbb{C}}(V)
$$

for some complex vector space $V$. A representation is called unitary if $V$ is a Hilbert space and the range of $\pi$ consists of unitary operators, and it is called continuous if $\pi$ is continuous.

Proposition $47 f \in B_{+}(G)$ if and only if there is a continuous unitary representation $\pi: G \rightarrow G L_{\mathbb{C}}(V)$ and a vector $v \in V$ such that

$$
\begin{equation*}
f=\langle\pi(\cdot) v, v\rangle . \tag{9}
\end{equation*}
$$

Proof $\Longleftarrow$ : For $f$ as in $(9), g_{1}, \ldots, g_{N} \in G$ and $c_{1}, \ldots, c_{N} \in \mathbb{C}$,

$$
\begin{aligned}
\sum_{i=1}^{N} \sum_{j=1}^{N} \overline{c_{i}} c_{j} f\left(g_{i}^{-1} g_{j}\right) & =\sum_{i=1}^{N} \sum_{j=1}^{N} \overline{c_{i}} c_{j}\left\langle\pi\left(g_{i}^{-1} g_{j}\right) v, v\right\rangle \\
& =\sum_{i=1}^{N} \sum_{j=1}^{N}\left\langle c_{j} \pi\left(g_{j}\right) v, c_{i} \pi\left(g_{i}\right) v\right\rangle \\
& =\left\langle\sum_{j=1}^{N} c_{j} \pi\left(g_{j}\right) v, \sum_{i=1}^{N} c_{i} \pi\left(g_{i}\right) v\right\rangle \geq 0
\end{aligned}
$$

using the fact that $\pi$ is a unitary representation. Moreover, $f$ is continuous and bounded since $\pi$ is, so $f \in B_{+}(G)$.
$\Longrightarrow$ : It can be obtained by combining Theorem 3.20 and Proposition 3.35 in Folland (2015).

Proposition $48 f \in B_{\mathbb{C}}(G)$ if and only if there is a continuous unitary representation $\pi: G \rightarrow G L_{\mathbb{C}}(V)$ and vectors $v, w \in V$ such that

$$
\begin{equation*}
f=\langle\pi(\cdot) v, w\rangle \tag{10}
\end{equation*}
$$

Proof $\Longrightarrow$ : We show that functions $f$ of the form (10) form a complex linear space. Closure under scalar multiplication can be obtained by scaling $v$. For closure under addition, observe that

$$
\left\langle\pi_{1}(\cdot) v_{1}, w_{1}\right\rangle+\left\langle\pi_{2}(\cdot) v_{2}, w_{2}\right\rangle=\left\langle\left(\pi_{1} \oplus \pi_{2}\right)(\cdot)\left(v_{1} \oplus v_{2}\right), w_{1} \oplus w_{2}\right\rangle
$$

and $\pi_{1} \oplus \pi_{2}$ is a continuous unitary representation if $\pi_{1}$ and $\pi_{2}$ are. Thus, functions of the form (10) form a complex linear space, and contain $B_{+}(G)$ by Proposition 47, so they contain $B_{\mathbb{C}}(G)$.
$\Longleftarrow$ : See (Kaniuth and Lau, 2018, Lemma 2.1.4).
We have described PD and PD decomposable functions on $G$ in the language of representation theory. What about their analogues on $H \backslash G / H$ ? From now on in this section we will view $C(H \backslash G / H)$ as a subset of $C(G)$. This will simplify the notation, and avoid confusing the symbol $\pi$ of the projection map with the representations, making the projection map implicit.

Lemma 49 For $f \in B_{+}(G)$,

$$
f=\langle\pi(\cdot) v, v\rangle
$$

as in (9), we have that $f \in C(H \backslash G / H)$ if and only if $\pi(h) v=v$ for all $h \in H$.
Proof $\Longleftarrow$ : If $\pi(h) v=v$ for all $h \in H$, then

$$
f\left(h g h^{\prime}\right)=\left\langle\pi\left(h g h^{\prime}\right) v, v\right\rangle=\left\langle\pi(g) \pi\left(h^{\prime}\right) v, \pi\left(h^{-1}\right) v\right\rangle=\langle\pi(g) v, v\rangle=f(g)
$$

for all $g \in G$ and $h, h^{\prime} \in H$. Thus, $f \in C(H \backslash G / H)$.

$$
\begin{aligned}
& \Longrightarrow: \text { If } f \in C(H \backslash G / H) \text {, for } h \in H, \\
& \qquad\langle\pi(h) v, \pi(h) v\rangle=\langle v, v\rangle=f(e)=f(h)=\langle\pi(h) v, v\rangle
\end{aligned}
$$

where the left most equality uses the fact that $\pi(h)$ is unitary. Thus,

$$
\langle\pi(h) v, v\rangle=\sqrt{\langle\pi(h) v, \pi(h) v\rangle\langle v, v\rangle}
$$

and by the equality condition of Cauchy-Schwarz's inequality, we deduce that $\pi(h) v=v$.
For $\pi$ a linear representation of $G$ into $G L_{\mathbb{C}}(V)$, write

$$
\begin{equation*}
W:=\{w \in V: \pi(h) w=w \forall h \in H\} \tag{11}
\end{equation*}
$$

and let $p: V \rightarrow W$ denote the orthogonal projection. In Lemma 49, we have shown $p(v)=v$ for the corresponding $\pi$ and $v$. For the more general setting $f \in B_{\mathbb{C}}(G)$, we do not necessarily have $p(v)=v$ or $p(w)=w$, but rather the following weaker lemma.

The proof for the forward direction of this lemma is technical and will require quite a bit of work to prove, so we move it to Appendix B. The reason for this difficulty is the generality of the setting we are working in, and specifically the generality of the locally compact group $G$. With such generality, there are few results that we can use about its continuous unitary representations. However, this will pay off with the generality that we shall obtain later on in Theorem 51.

Lemma 50 For $f \in B_{\mathbb{C}}(G)$,

$$
f=\langle\pi(\cdot) v, w\rangle
$$

as in (10), we have that $f \in C(H \backslash G / H)$ if and only if we can also write $f$ as

$$
f=\langle\pi(\cdot) p(v), p(w)\rangle
$$

with $p$ as defined above.
Proof $\Longleftarrow:$ If $g \in G$ and $h, h^{\prime} \in H$,

$$
f\left(h g h^{\prime}\right)=\left\langle\pi\left(h g h^{\prime}\right) p(v), p(w)\right\rangle=\left\langle\pi(g) \pi\left(h^{\prime}\right) p(v), \pi\left(h^{-1}\right) p(w)\right\rangle=\langle\pi(g) p(v), p(w)\rangle=f(g)
$$

so $f \in C_{b}(H \backslash G / H)$.
$\Longrightarrow$ : See Appendix B.
We are now in a position to prove our main theorem in this section.
Theorem 51 We have

1. $B_{\mathbb{R}}(G) \cap C(H \backslash G / H)=B_{\mathbb{R}}(H \backslash G / H)$,
2. $B_{\mathbb{C}}(G) \cap C(H \backslash G / H)=B_{\mathbb{C}}(H \backslash G / H)$.

Proof $\supseteq$ : For part 1 , if $f \in B_{\mathbb{R}}(H \backslash G / H)$ then $f \in C(H \backslash G / H)$ by definition, and $f$ has a PD decomposition, so $f \in B_{\mathbb{R}}(G)$ (or see the argument at the beginning of this section). Part 2 is similar.
$\subseteq$ : For part 1, let $f \in B_{\mathbb{R}}(G) \cap C(H \backslash G / H)$. By Proposition 47, there are continuous unitary representations $\pi_{ \pm}: G \rightarrow G L_{\mathbb{C}}\left(V_{ \pm}\right)$, and $v_{ \pm} \in V_{ \pm}$such that

$$
f=\left\langle\pi_{+}(\cdot) v_{+}, v_{+}\right\rangle-\left\langle\pi_{-}(\cdot) v_{-}, v_{-}\right\rangle=\left\langle\left(\pi_{+} \oplus \pi_{-}\right)(\cdot)\left(v_{+} \oplus v_{-}\right), v_{+} \oplus\left(-v_{-}\right)\right\rangle .
$$

Define $W_{ \pm} \leq V_{ \pm}$as

$$
W_{ \pm}:=\left\{w \in V_{ \pm}: \pi_{ \pm}(h) w=w \forall h \in H\right\}
$$

and $p_{ \pm}: V_{ \pm} \rightarrow W_{ \pm}$the orthogonal projections. For $W$ as in (11) and $p: V \rightarrow W$ the orthogonal projection, observe that $W=W_{+} \oplus W_{-}$and $p=p_{+} \oplus p_{-}$. Then by Lemma 50 we have

$$
\begin{aligned}
f & =\left\langle\left(\pi_{+} \oplus \pi_{-}\right)(\cdot) p\left(v_{+} \oplus v_{-}\right), p\left(v_{+} \oplus\left(-v_{-}\right)\right)\right\rangle \\
& =\left\langle\left(\pi_{+} \oplus \pi_{-}\right)(\cdot)\left(p_{+}\left(v_{+}\right) \oplus p_{-}\left(v_{-}\right)\right), p_{+}\left(v_{+}\right) \oplus p_{-}\left(-v_{-}\right)\right\rangle \\
& =\left\langle\pi_{+}(\cdot) p_{+}\left(v_{+}\right), p_{+}\left(v_{+}\right)\right\rangle-\left\langle\pi_{-}(\cdot) p_{-}\left(v_{-}\right), p_{-}\left(v_{-}\right)\right\rangle .
\end{aligned}
$$

Now $\left\langle\pi_{ \pm}(\cdot) p_{ \pm}\left(v_{ \pm}\right), p_{ \pm}\left(v_{ \pm}\right)\right\rangle \in B_{+}(H \backslash G / H)$ by Lemma 49, so $f \in B_{\mathbb{R}}(H \backslash G / H)$ and we are done.

For part 2 , the proof is similar by first writing $f$ as

$$
f=\left\langle\pi_{+}(\cdot) v_{+}, v_{+}\right\rangle-\left\langle\pi_{-}(\cdot) v_{-}, v_{-}\right\rangle+i\left\langle\pi_{+i}(\cdot) v_{+i}, v_{+i}\right\rangle-i\left\langle\pi_{-i}(\cdot) v_{-i}, v_{-i}\right\rangle .
$$

Remark 52 Note that the proof of Lemma 50 actually gives us the stronger results $B_{\mathbb{R}}(G) \cap$ $C(G / H)=B_{\mathbb{R}}(G / H)$ and $B_{\mathbb{C}}(G) \cap C(G / H)=B_{\mathbb{C}}(G / H)$.

### 5.2 Geometry and sufficient conditions for positive decomposability

Assuming some geometry on $G$ and $H$, we are now in a position to describe sufficient conditions for a function to be in $B_{\mathbb{C}}(H \backslash G / H)$. Specifically, we assume in this section that $G$ is a connected Lie group. Furthermore, we assume that $G$ is unimodular, meaning it possesses a Haar measure that is bi-invariant. This includes all Abelian and all compact Lie groups, but also many non-compact Lie groups of interest, such as $G L_{\mathbb{R}}(n)$ and $G L_{\mathbb{C}}(n)$.

Then, we have the following powerful characterization.
Theorem 53 Fix a left-invariant Riemannian metric on the connected unimodular Lie group $G$, and let $\Delta$ be the Laplace-Beltrami operator with respect to this metric. Suppose $f \in C^{\infty}(G)$ is such that

$$
\Delta^{k} f=\underbrace{\Delta \cdots \Delta}_{k \text { times }} f
$$

is $L^{2}$-integrable with respect to the Haar measure on $G$ for any integer $k \geq 0$. Then $f \in$ $B_{\mathbb{C}}(G)$.

Proof When $G$ is compact, the assumption can be reduced to $f \in C^{\infty}(G)$, and the result follows from Sugiura (1971), or from Gaudry and Pini (1986). For non-compact $G$, the result is a special case of (Gaudry et al., 1990, Theorem 1), which proves the inclusion of certain Besov spaces in Fourier algebras. See also the remark after (Gaudry et al., 1990,

Equation 12), as well as the containment of the Fourier algebra $A_{2}(G)$ in the FourierStieltjes algebra, for example in (Kaniuth and Lau, 2018, Proposition 2.3.3). We should note that the description of the Laplacian in Gaudry et al. (1990) as the sum of the squares of vector fields generating the Lie algebra $\mathfrak{g}$ of $G$ is non-standard, but we prove in Appendix C, that the Laplace-Beltrami operator arising from a left-invariant metric on a unimodular Lie group can be formulated in these terms.

Remark 54 For simplicity, Theorem 53 relaxes the results from the works referred to in its proof. For instance, by Gaudry and Pini (1986) we can show that when $G$ is compact, $C^{k}(G) \subset B_{\mathbb{C}}(G)$ for $k>n / 2$, where $n=\operatorname{dim} G$. This still does not capture the full strength of the result in this reference, which is a generalization of Bernstein's Theorem (see Example 41). On the other hand, when $G$ is non-compact, Gaudry et al. (1990) shows, in particular, that it suffices that $\Delta^{k} f$ is weakly in $L^{2}$ for $k \leq\left(n^{2}+n\right) / 4$ (see also (Varopoulos, 1988, Equation 2.6)).

Remark 55 Compare this with Example 41. In both cases, for a function $f$ to belong to $B_{\mathbb{C}}(G)$ it suffices that

1. $f$ is smooth enough,
2. $f$ and its derivatives decay fast enough at infinity.

We would now like to generalize Theorem 53 from $B_{\mathbb{C}}(G)$ to $B_{\mathbb{C}}(H \backslash G / H)$ with the help of Theorem 51. For this, we assume $H$ is a compact Lie subgroup of $G$. Then, $G / H$ is a homogeneous space, which inherits the structure of a manifold from the one on $G$. Moreover, observe that if the left-invariant Riemannian metric on $G$ is also right- $H$-invariant, then it gives rise to a natural Riemannian metric on the homogeneous space $G / H$, making the projection map $\pi_{R}: G \rightarrow G / H$ into a Riemannian submersion (see (Besse, 1987, Chapter $9)$ ). The ' $R$ ' in $\pi_{R}$ stands for 'right', and we will write $\pi_{L}: G / H \rightarrow H \backslash G / H$ for this other projection map, and $\pi=\pi_{L} \circ \pi_{R}: G \rightarrow H \backslash G / H$.

Remark 56 While $G / H$ inherits the structure of a manifold from the one on $G, H \backslash G / H$ does not. See Example 25 and Example 26.

Corollary 57 Suppose that the connected unimodular Lie group $G$ is equipped with a leftinvariant right- $H$-invariant Riemannian metric, where $H$ is a compact Lie subgroup of $G$. Equip $G / H$ with the Riemannian metric induced by the one on $G$. Suppose $\tilde{f} \in C^{\infty}(G / H)$ is such that there is $f \in C(H \backslash G / H)$ with $\tilde{f}=f \circ \pi_{L}$, i.e. $\tilde{f}$ is $H$-invariant. If $\Delta^{k} \tilde{f}$ is $L^{2}$-integrable with respect to the Haar measure on $G / H$ for any integer $k \geq 0$, then $f \in B_{\mathbb{C}}(H \backslash G / H)$.

Proof $f \in C(H \backslash G / H)$, so by Theorem 51, to show that $f \in B_{\mathbb{C}}(H \backslash G / H)$ we only need to prove that $f \circ \pi=\tilde{f} \circ \pi_{R} \in B_{\mathbb{C}}(G)$. Noting that compact groups are unimodular, we see by (Folland, 2015, Equation 2.52) that $\Delta^{k} \tilde{f}$ is $L^{2}$-integrable with respect to the Haar measure on $G / H$ if and only $\Delta^{k} \tilde{f} \circ \pi_{R}$ is $L^{2}$-integrable with respect to the Haar measure on $G$. Now, we show in Appendix C that $\left(\Delta^{k} \tilde{f}\right) \circ \pi_{R}=\Delta^{k}\left(\tilde{f} \circ \pi_{R}\right)$, so Theorem 53 tells us
that $\tilde{f} \circ \pi_{R} \in B_{\mathbb{C}}(G)$.

Example 58 By Corollary 57, any kernel $k=f(\langle\cdot, \cdot\rangle)$ that is a smooth function of the inner product has a PD decomposition on $S^{n}$ (see Example 25 for the description of functions on the double coset space of the sphere $S^{n} \cong S O(n+1) / S O(n)$, i.e. of invariant kernels on $\left.S^{n} \cong S O(n+1) / S O(n)\right)$. For instance, the hyperbolic tangent kernel $k=\tanh (a\langle\cdot, \cdot\rangle+b)$ on a sphere $S^{n}$, for $a, b \in \mathbb{R}$, has a PD decomposition, despite the fact that it is not PD in general (see (Smola et al., 2000, Example 5)).

We now consider the case where $M=G / H$ is a symmetric space of the non-compact type. We may choose $G$ and $H$ such that $G$ is connected, semisimple, has finite center and is noncompact, and $H$ is compact. The Riemannian metric on such a non-compact symmetric space is $G$-invariant, and always arises from a left-invariant right- $H$-invariant metric on $G$ (Helgason (1962)). In Corollary 59 and its proof below, we will use a shorthand and say a function "decays like" or "grows like" another function if it decays at least as fast or grows at most as fast as another function.

Corollary 59 Suppose $G / H$ is a non-compact semisimple symmetric space, with $G$ and $H$ chosen as above. Suppose $f \in C(H \backslash G / H)$ is a function of the Riemannian distance squared, i.e.

$$
f \circ \pi_{L}=\tilde{f}=g\left(d(e H, \cdot)^{2}\right)
$$

If $g \in C^{\infty}([0, \infty))$ is such that it and its derivatives decay exponentially at infinity, i.e.

$$
g^{(k)}(x)=o\left(e^{-c_{k} x}\right) \text { as } x \rightarrow \infty
$$

for all $k \geq 0$ and some $c_{k}>0$ possibly depending on $k$, then $f \in B_{\mathbb{C}}(H \backslash G / H)$. In particular, the Gaussian function $\exp \left(-\lambda d(e H, \cdot)^{2}\right) \in B_{\mathbb{R}}(H \backslash G / H)$, and the corresponding Gaussian kernel $k=\exp \left(-\lambda d(\cdot, \cdot)^{2}\right)$ has a $P D$ decomposition for any $\lambda>0$.

Proof This follows from Corollary 57 combined with two of Helgason's formulas for noncompact semisimple symmetric spaces, the first one for integration and the other for the radial part of the Laplace-Beltrami operator. Observe that since $G$ is semisimple, it is also unimodular, (Helgason, 1962, Chapter IV Proposition 1.4), so we are in a position to apply Corollary 57. Also note that since the metric on $G / H$ is left-invariant, any function of the distance squared $g(d(e H, \cdot))^{2}$ defines a function in $C(H \backslash G / H)$.

Using the notation from (Said et al., 2018, Equation 13), for $\tilde{f}$ an integrable function on $G / H$,

$$
\int_{G / H} \tilde{f}(x) d \nu(x)=C \int_{H} \int_{\mathfrak{a}} \tilde{f}(x(a, h)) D(a) d a d \mu(h)
$$

for some constant $C$ independent of $\tilde{f}$. Here $\mathfrak{a}$ is an Abelian subalgebra of $\mathfrak{g}$ resulting from the Iwasawa decomposition $\mathfrak{g}=\mathfrak{h}+\mathfrak{a}+\mathfrak{n}$ where $\mathfrak{g}, \mathfrak{h}$ are the Lie algebras of $G, H$ respectively, $\mathfrak{n}$ is a nilpotent subalgebra of $\mathfrak{g}, d a$ is the Lebesgue measure on $\mathfrak{a}, d \nu$ is a Haar measure on $G / H, d \mu$ is the normalised Haar measure of $H, x(a, h):=\exp (\operatorname{Ad}(h) a) H$ and

$$
D(a):=\prod_{\lambda>0} \sinh ^{m_{\lambda}}(|\lambda(a)|)
$$

where the product is over positive roots $\lambda: \mathfrak{a} \rightarrow \mathbb{R}$ and $m_{\lambda}$ is the dimension of the root space corresponding to $\lambda$. See (Helgason, 1962, Chapter IV Proposition 1.17). Few details will be important for us, only the form of the integral and the growth rate of $D$ at infinity.

Let $\tilde{f}$ and $g$ be as in the statement of the corollary. We first show that $\tilde{f}$ is $L^{2}$-integrable. We have

$$
\int_{G / H}|\tilde{f}(x)|^{2} d \nu(x)=A \int_{\mathfrak{a}}|\tilde{f}(x(a, e))|^{2} D(a) d a=A \int_{\mathfrak{a}}\left|g\left(\|a\|^{2}\right)\right|^{2} D(a) d a
$$

for some constant $A$, where we have noted that the integrand does not depend on $h$ and have integrated out that variable. The integrand is continuous, so to show that it is finite we only need to consider its behavior at infinity. Since $g$ decays exponentially and $D$ grows exponentially (by using that the roots are linear maps), the integrand decays like a Gaussian at infinity. We deduce that $\tilde{f}$ is $L^{2}$-integrable.

Now since $\tilde{f}$ is a function of distance only, $\Delta \tilde{f}=\Delta_{r} \tilde{f}$ where

$$
\begin{equation*}
\Delta_{r}=\sum_{i=1}^{r} \frac{\partial^{2}}{\partial a_{i}^{2}}+\sum_{\lambda>0} m_{\lambda} \operatorname{coth}(\lambda(a)) \frac{\partial}{\partial a_{\lambda}} \tag{12}
\end{equation*}
$$

is the radial part of the Laplace-Beltrami operator, where the $a_{i}$ form an orthonormal basis of $\mathfrak{a}$, the second sum is over positive roots $\lambda$, and $\lambda=B\left(a_{\lambda}, \cdot\right)$ where $B: \mathfrak{g} \times \mathfrak{g} \rightarrow \mathbb{R}$ is the Killing form. See (Helgason, 1984, Chapter II Theorem 5.24).

Since $g$ and $d(e H, \cdot)^{2}$ are smooth, so is $\tilde{f}$. Thus, to show that for $k \geq 0, \Delta^{k} \tilde{f}$ is $L^{2}$ integrable, it suffices to consider its behavior at infinity. It is not too hard to see from (12) that since $\tilde{f}(x(a, h))=g\left(\|a\|^{2}\right)=g\left(a_{1}^{2}+\cdots+a_{r}^{2}\right)$ and the derivatives of $g$ and of coth decay exponentially at infinity, $\Delta^{k} f$ decays like a Gaussian at infinity. Hence, arguing as above we deduce that $\Delta^{k} \tilde{f}$ is $L^{2}$-integrable. So we are done by Corollary 57 .

Remark 60 We can show the PD decomposability of the Gaussian kernel for reductive non-compact symmetric spaces too. For instance, the space of $n \times n$ symmetric positivedefinite matrices $\mathbb{S}_{++}^{n} \cong G L(n) / O(n)$ is not a semisimple symmetric space. However, we can also write it as $\mathbb{S}_{++}^{n} \cong S L(n) / S O(n) \times \mathbb{R}_{>0}$, with the diffeomorphism

$$
\begin{align*}
\mathbb{S}_{++}^{n} & \rightarrow S L(n) / S O(n) \times \mathbb{R}_{>0}  \tag{13}\\
X & \mapsto\left(X / \operatorname{det}(X)^{n}, \operatorname{det}(X)\right)
\end{align*}
$$

where $S L(n) / S O(n)$ is viewed as the space of symmetric positive-definite matrices with unit determinant. Then, if we write $d$ for the Riemannian distance on $\mathbb{S}_{++}^{n} \cong G L(n) / O(n)$, sometimes called the affine-invariant Riemannian distance, and $d^{\prime}$ for the Riemannian distance on $S L(n) / S O(n)$, we can view the map (13) as an isometry:

$$
d(X, Y)=d^{\prime}\left(X / \operatorname{det}(X)^{n}, Y / \operatorname{det}(Y)^{n}\right)+|\log (\operatorname{det}(X))-\log (\operatorname{det}(Y))|
$$

for all $X, Y \in \mathbb{S}_{++}^{n}$. Therefore, the Gaussian kernel on $\mathbb{S}_{++}^{n}$ can be viewed as
$\exp \left(-\lambda d(X, Y)^{2}\right)=\exp \left(-\lambda d^{\prime}\left(X / \operatorname{det}(X)^{n}, Y / \operatorname{det}(Y)^{n}\right)^{2}\right) \exp \left(-\lambda|\log (\operatorname{det}(X))-\log (\operatorname{det}(Y))|^{2}\right)$
a product of a PD decomposable kernel by Corollary 59 with a PD kernel, which is PD decomposable by Example 18.

## 6 Conclusion

We have seen that kernel methods can be applied with non-PD kernels that admit a PD decomposition and that knowledge of the specific form of the decomposition is not necessary for RKKS learning. We have argued that RKKS-based methods are particularly interesting in non-Euclidean settings where PD kernels are challenging to construct. We have then related PD decomposable invariant kernels on locally compact quotient spaces $G / H$ to Hermitian PD decomposable functions on the double coset space $H \backslash G / H$. This allowed us to leverage the extensive harmonic analytic literature on the Fourier-Stieltjes algebra $B_{\mathbb{C}}(G)$ to describe the invariant kernels that admit a PD decomposition. In particular, assuming some geometry on $G / H$, we showed that smoothness and appropriate decay of the derivatives at infinity is sufficient for the PD decomposability of a kernel, providing weak and verifiable sufficient conditions for a kernel to admit a PD decomposition. This work provides a theoretical foundation for applications of kernel methods on non-Euclidean data spaces.

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## Appendix A.

This appendix deals with two problems: firstly, we show that when $G$ is compact, the correspondence from Proposition 29 between Hermitian functions on $H \backslash G / H$ with continuous PD decompositions and kernels on $G / H$ with continuous PD decompositions is a surjection. As explained in Remark 31, this requires PD decomposable invariant kernels to admit invariant PD decompositions. Secondly, we show that when $H$ is compact, a PD decomposable function $f$ on $G$ that is invariant on double cosets has a PD decomposition that is invariant on double cosets. This is a special case of Theorem 51 from Section 5.1, but assuming compactness of $H$ gives a significantly shorter proof.

These two results are grouped together in this appendix because their proofs are in spirit the same.

Theorem If the locally compact group $G$ is compact and $H$ is a closed subgroup of $G$, then an invariant kernel $k$ on $G / H$ has a PD decomposition if and only if it has a PD decomposition into invariant kernels.

Proof For $g \in G$, define $T_{g}$ to be the operator acting on maps $k: G / H \times G / H \rightarrow \mathbb{C}$ by

$$
T_{g} k(a H, b H)=k(g a H, g b H)
$$

for all $a, b \in G$. Let $k$ be an invariant PD decomposable kernel on $G / H$. Then, by invariance of $k$,

$$
k(a H, b H)=\frac{1}{\mu(G)} \int_{G} T_{g} k(a H, b H) d \mu(g)
$$

for $a, b \in G$, where $\mu$ is a Haar measure for $G$, which is finite by compactness of $G$. So writing the PD decomposition of $k$ as $k=k_{+}-k_{-}$, we get

$$
k=\frac{1}{\mu(G)} \int_{G} T_{g} k_{+} d \mu(g)-\frac{1}{\mu(G)} \int_{G} T_{g} k_{-} d \mu(g) .
$$

Both the first term and minus the second term are PD, and they are invariant by leftinvariance of the Haar measure, so this is a PD decomposition into invariant kernels.

We now prove in a similar way the following special case to Theorem 51, writing for simplicity $C(H \backslash G / H) \subset C(G)$ as in Section 5.1:

Theorem If the closed subgroup $H$ of the locally compact group $G$ is compact, then $f \in$ $B_{\mathbb{R}}(G) \cap C(H \backslash G / H)$ implies $f \in B_{\mathbb{R}}(H \backslash G / H)$. We have an analogous statement for $B_{\mathbb{C}}(G)$ and $B_{\mathbb{C}}(H \backslash G / H)$.

Proof For $g \in G$, define $L_{g}$ and $R_{g}$ to be the operators acting on functions $f: G \rightarrow \mathbb{C}$ by

$$
L_{g} f(a)=f(g a)
$$

and

$$
R_{g} f(a)=f(a g)
$$

for all $a \in G$. Suppose $f \in B_{\mathbb{R}}(G) \cap C(H \backslash G / H)$. Then, by invariance of $f$ on double cosets,

$$
f(g)=\frac{1}{\nu(H)^{2}} \int_{H} \int_{H} R_{h^{\prime}} L_{h} f(g) d \nu(h) d \nu\left(h^{\prime}\right)
$$

for $g \in G$, where $\nu$ is a translation bi-invariant Haar measure on $H$, which exists and is finite by compactness of $H$. So writing the PD decomposition of $f$ as $f=f_{+}-f_{-}$, we get

$$
f=\frac{1}{\nu(H)^{2}} \int_{H} \int_{H} R_{h^{\prime}} L_{h} f_{+} d \nu(h) d \nu\left(h^{\prime}\right)-\frac{1}{\nu(H)^{2}} \int_{H} \int_{H} R_{h^{\prime}} L_{h} f_{-} d \nu(h) d \nu\left(h^{\prime}\right) .
$$

Both the first term and minus the second term are PD, and they are invariant on double cosets by bi-invariance of the Haar measures. Continuity of the terms is easy to check, and the proof for $B_{\mathbb{C}}(G)$ and $B_{\mathbb{C}}(H \backslash G / H)$ is similar.

## Appendix B.

In this appendix, we give the proof of the forward direction of Lemma 50. Recall the statement of this lemma:

Lemma For $f \in B_{\mathbb{C}}(G)$, write

$$
f=\langle\pi(\cdot) v, w\rangle
$$

where $\pi: G \rightarrow G L_{\mathbb{C}}(V)$ is a continuous unitary representation and $V$ is a Hilbert space. Then we have that $f \in C(H \backslash G / H)$ if and only if we can also write $f$ as

$$
f=\langle\pi(\cdot) p(v), p(w)\rangle
$$

where $p: V \rightarrow W$ is the orthogonal projection and $W:=\{w \in V: \pi(h) w=w \forall h \in H\}$.
Proof $\Longrightarrow$ : fix $g \in G$ and consider the set

$$
\begin{aligned}
\mathcal{P}:= & \left\{Z: W \leq Z \leq V \text { s.t. } Z=\bar{Z} \text { and }\langle\pi(g) v, w\rangle=\left\langle\pi\left(h g h^{\prime}\right) v, q(w)\right\rangle\right. \\
& \left.\forall h, h^{\prime} \in H \text { where } q: V \rightarrow Z \text { is the orthogonal projection }\right\} .
\end{aligned}
$$

$\mathcal{P}$ is partially ordered by inclusion. We want to apply Zorn's lemma to $\mathcal{P}$. So we break down our proof into two steps: we first show in step I that every chain-i.e. totally ordered subset - of $\mathcal{P}$ has a lower bound in $\mathcal{P}$. Zorn's lemma then tells us that $\mathcal{P}$ has a minimal element. Then we show in step II that this minimal element is $W$. I may seem obvious but requires some care, since a chain may be uncountable and need not be well-ordered. II on the other hand will require us to look more closely at the representation $\pi$. We will expand it in irreducible components along a cyclic subgroup of $H$.
$I:$ Let $\mathcal{C} \subset \mathcal{P}$ be a chain. We want to show that $\mathcal{C}$ has a lower bound in $\mathcal{P}$. This lower bound will of course be $Y:=\bigcap_{Z \in \mathcal{C}} Z$, so we only need to show $Y \in \mathcal{P}$. Note that for $Z \in \mathcal{C}, Y \subset Z$ so $Y^{\perp} \supset Z^{\perp}$, thus taking finite sums of elements and closure over such $Z$ we get $Y^{\perp} \supset \overline{\sum_{Z \in \mathcal{C}} Z^{\perp}}$. We can also see that for $Z^{\prime} \in \mathcal{C}, Z^{\prime} \supset\left(\sum_{Z \in \mathcal{C}} Z^{\perp}\right)^{\perp}$, so taking
the intersection over such $Z^{\prime}$ we get $Y \supset\left(\sum_{Z \in \mathcal{C}} Z^{\perp}\right)^{\perp}$ and taking orthogonal complements $Y^{\perp} \subset \overline{\sum_{Z \in \mathcal{C}} Z^{\perp}}$. So we have shown

$$
\begin{equation*}
Y^{\perp}=\overline{\sum_{Z \in \mathcal{C}} Z^{\perp}} \tag{14}
\end{equation*}
$$

Let $r: V \rightarrow Y, r^{\perp}: V \rightarrow Y^{\perp}$ be the orthogonal projections, and for $Z \in \mathcal{C}$, let $q_{Z}: V \rightarrow Z$, $q_{Z}^{\perp}: V \rightarrow Z^{\perp}$ be the orthogonal projections. We have for such $Z$ and $h, h^{\prime} \in H$,

$$
\left\langle\pi\left(h g h^{\prime}\right) v, q_{Z}(w)\right\rangle=\langle\pi(g) v, w\rangle=\left\langle\pi\left(h g h^{\prime}\right) v, w\right\rangle=\left\langle\pi\left(h g h^{\prime}\right) v, q_{Z}(w)\right\rangle+\left\langle\pi\left(h g h^{\prime}\right) v, q_{Z}^{\frac{1}{Z}}(w)\right\rangle
$$

so $\left\langle\pi\left(h g h^{\prime}\right) v, q_{Z}^{\perp}(w)\right\rangle=0$. Similarly, to show $Y \in \mathcal{P}$, we only need to show $\left\langle\pi\left(h g h^{\prime}\right) v, r^{\perp}(w)\right\rangle=$ 0 . Now by (14), $r^{\perp}(w)$ can be written as a series of countably many elements of the $Z^{\perp}$ for $Z \in \mathcal{C}$, so $q_{Z}^{\perp}\left(r^{\perp}(w)\right)$ gets arbitrarily close to $r^{\perp}(w)$ as $Z$ decreases (we don't use the notion of limit here as $\mathcal{C}$ may be uncountable). Now $q_{Z}^{\perp}(w)=q_{Z}^{\perp}\left(r^{\perp}(w)\right)$, so $\left\langle\pi\left(h g h^{\prime}\right) v, q_{Z}^{\perp}(w)\right\rangle=0$ for all $Z \in \mathcal{C}$ implies $\left\langle\pi\left(h g h^{\prime}\right) v, r^{\perp}(w)\right\rangle=0$, and we are done.
$I I$ : By Zorn's lemma, $\mathcal{P}$ has a minimal element, which we denote by $Z$ and write $q: V \rightarrow Z$ for the corresponding orthogonal projection. If $q(w)=p(w)$, then by minimality of $Z$ we have $Z=W$. Otherwise, suppose first $q(w) \neq p(w)$. Then there is $h \in H$ such that $\pi(h) q(w) \neq q(w)$. Let $C_{h}$ be the cyclic group generated by $h$. For conciseness we will treat the cases $C_{h}$ finite and $C_{h}$ infinite together, and only use the fact that $C_{h}$ is Abelian. Indeed, (Folland, 2015, Theorem 4.45) tells us that we can decompose $\left.\pi\right|_{C_{h}}$ into irreducible components in the form

$$
\pi\left(h^{k}\right)=\int_{\widehat{C}_{h}} \xi\left(h^{k}\right) d P(\xi)
$$

for all $k \in \mathbb{Z}$, for some regular $V$-projection-valued measure $P$ on $\widehat{C}_{h}$ (see (Folland, 2015, Theorem 1.38) and the following discussion for the definition of a projection-valued measure). Then

$$
\begin{aligned}
\langle\pi(g) v, w\rangle & =\left\langle\pi\left(h^{\prime} g h^{\prime \prime}\right) v, \pi\left(h^{k}\right) q(w)\right\rangle \\
& =\left\langle\pi\left(h^{\prime} g h^{\prime \prime}\right) v, \int_{\widehat{C}_{h}} \xi\left(h^{k}\right) d P(\xi) q(w)\right\rangle \\
& =\int_{\widehat{C}_{h}} \xi\left(h^{k}\right) d P_{\pi\left(h^{\prime} g h^{\prime \prime}\right) v, q(w)}(\xi) \\
& =\mathscr{F}^{-1} P_{\pi\left(h^{\prime} g h^{\prime \prime}\right) v, q(w)}\left(h^{k}\right)
\end{aligned}
$$

for all $k \in \mathbb{Z}$, where $P_{\pi\left(h^{\prime} g h^{\prime \prime}\right) v, q(w)}$ is a finite complex Radon measure, i.e. $P_{\pi\left(h^{\prime} g h^{\prime \prime}\right) v, q(w)} \in$ $M_{\mathbb{C}}\left(\widehat{C}_{h}\right)$, and $\mathscr{F}^{-1}$ is the inverse Fourier transform. Then by injectivity of $\mathscr{F}^{-1}$ we must have

$$
P_{\pi\left(h^{\prime} g h^{\prime \prime}\right) v, q(w)}=\langle\pi(g) v, w\rangle \delta_{1}
$$

where $\delta_{1}$ is the Dirac measure at $\xi=1$.
Now consider

$$
Y:=\{z \in Z: \pi(h) z=z\}
$$

where $h$ is as before. So for all $h^{\prime}, h^{\prime \prime} \in H$,

$$
\begin{aligned}
\langle\pi(g) v, w\rangle & =P_{\pi\left(h^{\prime} g h^{\prime \prime}\right) v, q(w)}(1) \\
& =\int_{\{1\}} d P_{\pi\left(h^{\prime} g h^{\prime \prime}\right) v, q(w)}(\xi) \\
& =\left\langle\pi\left(h^{\prime} g h^{\prime \prime}\right) v, \int_{\{1\}} d P(\xi) q(w)\right\rangle \\
& =\left\langle\pi\left(h^{\prime} g h^{\prime \prime}\right) v, P(1) q(w)\right\rangle \\
& =\left\langle\pi\left(h^{\prime} g h^{\prime \prime}\right) v, r(w)\right\rangle
\end{aligned}
$$

where $r: V \rightarrow Y$ is the orthogonal projection. Hence, $Y \in \mathcal{P}$. But $\pi(h) q(w) \neq q(w)$ so $r(w) \neq q(w)$, which implies $Y<Z$. This contradicts minimality of $Z$. Thus, $Z=W$, and hence $W \in \mathcal{P}$.

Now by noting that

$$
\left\langle\pi\left(h g h^{\prime}\right) v, p(w)\right\rangle=\left\langle q(v), \pi\left(\left(h g h^{\prime}\right)^{-1}\right) p(w)\right\rangle=\overline{\left\langle\pi\left(h^{\prime-1} g^{-1} h^{-1}\right) p(w), v\right\rangle},
$$

for all $h, h^{\prime} \in H$, we can apply the same argument on $v$ to obtain

$$
\langle\pi(g) v, w\rangle=\left\langle\pi\left(h^{\prime} g h^{\prime \prime}\right) p(v), p(w)\right\rangle
$$

for all $h^{\prime}, h^{\prime \prime} \in H$. Finally, $g \in G$ was arbitrary, so we are done.

## Appendix C.

In this appendix we investigate properties of the Laplace-Beltrami operator on the Lie group $G$, which is equipped with a left-invariant Riemannian metric. We will use the following definition of the Laplace-Beltrami operator:

$$
\Delta:=\sum_{i=1}^{n}\left(X_{i}^{2}-\nabla_{X_{i}} X_{i}\right)
$$

locally around a point $g \in G$, where $X_{1}, \ldots, X_{n}$ is a local orthonormal frame of vector fields around $g$ and $\nabla$ is the Levi-Civita connection with respect to the metric on $G$. This definition of $\Delta$ is independent of the choice of local orthonormal frame.

In the proof of Theorem 53, we implicitly use the following:
Proposition Let $G$ be a connected unimodular Lie group. If $X_{1}, \ldots, X_{n}$ is a basis of leftinvariant vector fields for $G$, and we equip $G$ with the left-invariant Riemannian metric making $X_{1}, \ldots, X_{n}$ orthonormal, then the Laplace-Beltrami operator $\Delta$ can be written as

$$
\Delta=\sum_{i=1}^{n} X_{i}^{2}
$$

Proof We have

$$
\Delta=\sum_{i=1}^{n}\left(X_{i}^{2}-\nabla_{X_{i}} X_{i}\right)
$$

so we need to show that $\sum_{i=1}^{n} \nabla_{X_{i}} X_{i}=0$.
$G$ being unimodular is equivalent to the adjoint representation ad : $\mathfrak{g} \rightarrow \operatorname{Lie}(\operatorname{Aut}(\mathfrak{g}))$ of the Lie algebra $\mathfrak{g}$ of $G$ having vanishing trace at every $X \in \mathfrak{g}$ (Milnor, 1976, Lemma 6.3), where $\operatorname{Lie}(\operatorname{Aut}(\mathfrak{g}))$ is the Lie algebra of $\operatorname{Aut}(\mathfrak{g})$, the Lie group of linear automorphisms of $\mathfrak{g}$. So for all $i$,

$$
\begin{aligned}
0 & =\operatorname{tr}\left(\operatorname{ad}\left(X_{i}\right)\right) \\
& =\sum_{j=1}^{n}\left\langle\left[X_{i}, X_{j}\right], X_{j}\right\rangle_{\mathfrak{g}} \\
& =\sum_{j=1}^{n}(\underbrace{\left\langle\nabla_{X_{i}} X_{j}, X_{j}\right\rangle_{\mathfrak{g}}}_{=0}-\left\langle\nabla_{X_{j}} X_{i}, X_{j}\right\rangle_{\mathfrak{g}}) \\
& =-\sum_{j=1}^{n}\left\langle\nabla_{X_{j}} X_{i}, X_{j}\right\rangle_{\mathfrak{g}} \\
& =\sum_{j=1}^{n}\left\langle\nabla_{X_{j}} X_{j}, X_{i}\right\rangle_{\mathfrak{g}} \\
& =\left\langle\sum_{j=1}^{n} \nabla_{X_{j}} X_{j}, X_{i}\right\rangle_{\mathfrak{g}}
\end{aligned}
$$

where we used orthonormality of the $X_{j}$ in the fourth and fifth equality. So $\sum_{i=1}^{n} \nabla_{X_{i}} X_{i}=0$ at $e$, and thus everywhere on $G$ by left-invariance.
Now in the proof of Corollary 57, we need the following:
Proposition Let $G$ be a Lie group and let $H$ be a compact Lie subgroup of $G$. Suppose $G$ is equipped with a left-invariant, right- $H$-invariant Riemannian metric. Equip $G / H$ with the Riemannian metric induced by the one on $G$. Then, for $\tilde{f} \in C^{\infty}(G / H)$,

$$
\Delta\left(\tilde{f} \circ \pi_{R}\right)=\Delta \tilde{f} \circ \pi_{R}
$$

where $\pi_{R}: G \rightarrow G / H$ is the projection, and $\Delta$ is the Laplace-Beltrami operator with respect to the respective metrics.

Proof This follows many of the steps of (Bergery and Bourguignon, 1982, proof of Theorem 1.5). Write $n$ and $m$ for the dimensions of $G$ and $H$ respectively. For $g \in G$, take a local orthonormal frame $\tilde{X}_{1}, \ldots, \tilde{X}_{n-m}$ for $T_{g H} G / H$ around $g H$. This lifts to a local orthonormal frame $X_{1}, \ldots, X_{n-m}$ for $\operatorname{Ker}\left(\pi_{R *}\right)^{\perp}$ around $g$, subbundle of $T G$. So $\pi_{R *} X_{i}=\tilde{X}_{i}$ for all $i$. Also take a local orthonormal frame $U_{1}, \ldots, U_{m}$ for $\operatorname{Ker}\left(\pi_{R *}\right)$ around $g$. Then, the LaplaceBeltrami operator on $G$ can be locally written as

$$
\Delta=\sum_{i=1}^{n-m}\left(X_{i}^{2}-\nabla_{X_{i}} X_{i}\right)+\sum_{i=1}^{m}\left(U_{i}^{2}-\nabla_{U_{i}} U_{i}\right)
$$

and the Laplace-Beltrami operator on $G / H$ can be locally written as

$$
\Delta=\sum_{i=1}^{n-m}\left(\tilde{X}_{i}^{2}-\nabla_{\tilde{X}_{i}} \tilde{X}_{i}\right)
$$

where $\nabla$ are the Levi-Civita connections induced by the respective metrics. By the chain rule, $U\left(\tilde{f} \circ \pi_{R}\right)=\pi_{R *} U(\tilde{f}) \circ \pi_{R}=0$ for all $U \in \operatorname{Ker}\left(\pi_{R *}\right)$. Similarly, the chain rule gives $X_{i}^{2}\left(\tilde{f} \circ \pi_{R *}\right)=\pi_{R *} X_{i} \pi_{R *} X_{i}(\tilde{f}) \circ \pi_{R}=\tilde{X}_{i}^{2}(\tilde{f}) \circ \pi_{R}$ for all $i$. Now by (Besse, 1987, Theorem 9.80) if $U \in \operatorname{Ker}\left(\pi_{R *}\right)$ then $\nabla_{U} U \in \operatorname{Ker}\left(\pi_{R *}\right)$ (we use here the left-invariance and right- $H$-invariance of the metric on $G$, and the compactness of $H$ ). Finally, by (Besse, 1987, Definition 9.23 \& following discussion), $\nabla_{X_{i}} X_{i}\left(\tilde{f} \circ \pi_{R}\right)=\pi_{R *}\left(\nabla_{X_{i}} X_{i}\right)(\tilde{f}) \circ \pi_{R}=$ $\nabla_{\tilde{X}_{i}} \tilde{X}_{i}(\tilde{f}) \circ \pi_{R}$ for all $i$. Thus, we obtain

$$
\begin{aligned}
\Delta\left(\tilde{f} \circ \pi_{R}\right) & =\sum_{i=1}^{n-m}\left(X_{i}^{2}-\nabla_{X_{i}} X_{i}\right)\left(\tilde{f} \circ \pi_{R}\right)+\sum_{i=1}^{m}\left(U_{i}^{2}-\nabla_{U_{i}} U_{i}\right)\left(\tilde{f} \circ \pi_{R}\right) \\
& =\sum_{i=1}^{n-m}\left(\tilde{X}_{i}^{2}-\nabla_{\tilde{X}_{i}} \tilde{X}_{i}\right)(\tilde{f}) \circ \pi_{R} \\
& =\Delta \tilde{f} \circ \pi_{R} .
\end{aligned}
$$

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