

# Derivatives of Markov kernels and their Jordan decomposition

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## Abstract

We study a particular class of transition kernels that stems from differentiating Markov kernels in the weak sense. Sufficient conditions are established for this type of kernels to admit a Jordan–type decomposition. The decomposition is explicitly constructed.

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# 1 Introduction

Let  $P_\theta$  be a family of Markov kernels from a measurable space  $(X, \mathcal{X})$  to a locally compact space  $Y$  (a precise definition will be given later in the text), with  $\theta \in \Theta \subset \mathbb{R}$ , and let  $\mathcal{C}_c(Y)$  denote the set of continuous real-valued mappings with compact support on  $Y$ . The Markov kernel  $P_\theta$  is called *weakly differentiable* at  $\theta$  if for any  $x \in X$  a finite signed measure  $P'_\theta(x; \cdot)$  on  $(Y, \mathcal{Y})$  exists such that for any  $g \in \mathcal{C}_c(Y)$ :

$$\frac{d}{d\theta} \int g(y) P_\theta(x; dy) = \int g(y) P'_\theta(x; dy). \quad (1)$$

This definition of weak differentiability is slightly more general than the original one in [4]: there (1) has to hold for any continuous bounded mapping  $g$ . Weak differentiability has been successfully applied to the theory of Markov chains. See [1] for an application to a problem in maintenance theory and [2] for an application to option pricing. The concept of weak differentiation is also related to finding optimal statistical tests, see [7]. For Markov chains, the following result is of particular interest: let  $\pi_\theta$  denote the (unique) invariant distribution of  $P_\theta$  (existence is assumed here), then it can be shown that

$$\pi'_\theta = \pi_\theta \sum_{n=0}^{\infty} P'_\theta P_\theta^n, \quad (2)$$

where  $P'_\theta$  is defined through (1) and  $P_\theta^n$  denotes the  $n$  fold product of  $P_\theta$ , see [4, 3] for a proof and more details on weak differentiability. If  $P'_\theta$  exists, then the fact that  $P'_\theta(x; \cdot)$  fails to be a probability measure poses the problem of sampling from  $P'_\theta$ . For  $x \in X$  fixed, we can represent  $P'_\theta(x; \cdot)$  by its Jordan decomposition as a difference between two probability measures as follows. For a finite signed measure  $\mu$  denote its Jordan decomposition by  $[\mu]^+$  and  $[\mu]^-$ , i.e.,  $\mu = [\mu]^+ - [\mu]^-$  and  $[\mu]^+, [\mu]^-$  are positive measures. Let

$$c_{P_\theta}(x) = [P'_\theta]^+(x; X) = [P'_\theta]^-(x; X) \quad (3)$$

and

$$P_\theta^+(x; \cdot) = \frac{[P'_\theta]^+(x; \cdot)}{c_{P_\theta}(x)}, \quad P_\theta^-(x; \cdot) = \frac{[P'_\theta]^-(x; \cdot)}{c_{P_\theta}(x)},$$

then it holds, for all  $g \in \mathcal{C}_c(Y)$ , that

$$\int g(y) P'_\theta(x; dy) = c_{P_\theta}(x) \left( \int g(y) P_\theta^+(x; dy) - \int g(y) P_\theta^-(x; dy) \right). \quad (4)$$

For the above line of argument we fixed  $x$ . For  $P_\theta^+$  and  $P_\theta^-$  to be Markov kernels, we have to consider  $P_\theta^+$  and  $P_\theta^-$  as functions in  $x$  and have to establish

measurability of  $P_\theta^+(\cdot; A)$  and  $P_\theta^-(\cdot; A)$  for any  $A \in \mathcal{Y}$ . The solution of this problem implies that  $c_{P_\theta}(\cdot)$  in (3) is measurable as a mapping from  $X$  to  $\mathbb{R}$ . A representation of  $P'_\theta$  through  $(c_{P_\theta}(\cdot), P_\theta^+, P_\theta^-)$ , with  $c_{P_\theta}$  measurable and  $P_\theta^\pm$  Markov kernels, is called a *weak derivative* of  $P_\theta$ . The existence of a weak derivative is of key importance for the statistical interpretation of (2) and for obtaining efficient unbiased gradient estimators.

In this paper, we give sufficient conditions for  $P'_\theta$  to possess a representation as scaled difference of two Markov kernels. Specifically, we show that uniform boundedness of  $P'_\theta$  ( i.e., the supremum of  $|\int g(y)P_\theta(x; dy)|$  over  $g \in \mathcal{C}_c(Y)$  with  $|g| \leq 1$  and  $x \in X$  is finite) is together with a topological condition on  $Y$  sufficient for  $c_{P_\theta}(\cdot)$  in (3) to be measurable (and for  $P_\theta^+$  and  $P_\theta^-$  to be Markov kernels again). In conclusion we will show that uniform boundedness is sufficient for  $P'_\theta$  to admit a weak derivative.

The paper is organized as follows. Section 1 introduces the basic concepts and definitions. Section 2 shows that, under suitable conditions, the kernel  $P'_\theta$  as defined in (1) can be uniquely extended to the bounded Borel-measurable mappings. In Section 3 an explicit construct of a Jordan-type decomposition of  $P'_\theta$  is given.

## 2 Conditional Integrals and Kernels

We say that a topological space is *second countable* if its topology is generated by a countable basis, i.e., if there exists a countable family of open (or closed) sets which generates the topology. Throughout the paper we let  $Y$  always denote a locally compact second countable Hausdorff space. We denote by  $\mathcal{Y}$  the  $\sigma$ -field of Baire measurable subsets of  $Y$ , i.e., the  $\sigma$ -field generated by the compact subsets of  $Y$ .

**Remark 1** *On a second countable locally compact space the Borel-field (the  $\sigma$ -field generated by the open or closed sets) and the Baire-field coincide. (This holds true since any open set in a second countable locally compact space is a countable union of compact sets.) Thus,  $\mathcal{Y}$  is the  $\sigma$ -field generated by the family of open sets in  $Y$ .*

*For example, the space  $\mathbb{R}^n$  and any submanifold of it constitutes a locally compact second countable space.*

**Remark 2** *Notice that a metrizable space is second countable if and only if it is separable (see [8] Theorem 16.11). Conversely, a locally compact or even a compact space may be separable but not second countable. An example of*

a separable compact space that fails to be second countable is provided by the Stone-Cech compactification of the natural numbers.

Let  $X$  be an arbitrary set and let  $\mathcal{X}$  be an arbitrary  $\sigma$ -field on  $X$ . Let  $\mathcal{B}_b(Y)$  be the family of real-valued bounded  $\mathcal{Y}$ -measurable functions on  $Y$ , let  $\mathcal{C}_c$  the family of continuous functions with compact support on  $Y$  and let  $\mathcal{B}(X)$  denote the family of real-valued  $\mathcal{X}$ -measurable functions on  $X$ .

We call a Baire measurable function, say  $g$ , *simple* if and only if an integer  $n \in \mathbb{N}$  and, for  $i \leq n$ , sets  $B_i \in \mathcal{Y}$  and constants  $\gamma_i \in \mathbb{R}$  exist such that

$$g(y) = \sum_{i=1}^n \gamma_i \mathbf{1}_{B_i}(y), \quad y \in Y.$$

The family of Baire measurable simple functions on  $Y$  is denoted by  $\mathcal{B}_{simp}(Y)$ .

We note that  $\mathcal{C}_c(Y) \subset \mathcal{B}_b(Y)$  and define the supremum norm  $\|\cdot\|$  on  $\mathcal{B}_b(Y)$  by

$$\|g\| := \sup_{y \in Y} |g(y)|.$$

We call a set  $\mathcal{G} \subset \mathcal{B}_b(Y)$  *uniformly bounded* or *sup-norm bounded* if

$$\sup_{g \in \mathcal{G}} \|g\| < \infty.$$

We say that a sequence  $(g_n)_{n \in \mathbb{N}}$  of functions  $g_n \in \mathcal{B}_b(Y)$  is uniformly bounded if the set  $\{g_n \mid n \in \mathbb{N}\}$  is uniformly bounded.

We say that a linear functional  $J : \mathcal{C}_c(Y) \rightarrow \mathbb{R}$  is an *integral* if it is bounded on uniformly bounded subsets of  $\mathcal{C}_c(Y)$  (such functionals may also be called sup-norm bounded). We say that a linear functional  $\tilde{J} : \mathcal{B}_b(Y) \rightarrow \mathbb{R}$  is an *extended integral* if it is bounded on uniformly bounded subsets  $\mathcal{G}$  of  $\mathcal{B}_b(Y)$ .

We say that a sequence  $(f_n)_{n \in \mathbb{N}}$  of functions  $f_n$  from some set  $S$  to a Hausdorff space  $V$  *converges point-wise* if  $\lim_{n \rightarrow \infty} f_n(s)$  exists for any  $s \in S$ .

**Definition 1** A kernel  $P(\cdot, \cdot)$  from  $X$  to  $Y$  is a function  $P : X \times \mathcal{Y} \rightarrow \mathbb{R}$  such that  $P(x, \cdot)$  is for any  $x \in X$  a finite signed measure on  $(Y, \mathcal{Y})$  and  $x \mapsto P(x, B)$  is for any  $B \in \mathcal{Y}$  a  $\mathcal{X}$ -measurable function on  $X$ . We say that the kernel is *Markov* (or a *Markov kernel*) if for any  $x \in X$  the measure  $P(x, \cdot)$  is a probability measure. We denote the space of all kernels from  $X$  to  $Y$  by  $\mathcal{P}(X, Y)$ .

**Definition 2** A conditional integral  $I(\cdot, \cdot)$  from  $X$  to  $\mathcal{C}_c(Y)$  is a function  $I : X \times \mathcal{C}_c(Y) \rightarrow \mathbb{R}$  such that

- $I(x, \cdot)$  is an integral (i.e. a linear functional on  $\mathcal{C}_c(Y)$  which is sup-norm bounded) and
- $x \mapsto I(x, f)$  is for any  $f \in \mathcal{C}_c(Y)$  a  $\mathcal{X}$ -measurable function on  $X$ .

We denote the space of conditional integrals from  $X$  to  $\mathcal{C}_c(Y)$  by  $\mathcal{I}(X, Y)$ .

**Definition 3** Let  $Z$  denote an arbitrary Hausdorff space. We say that a function  $F : \mathcal{B}_b(Y) \mapsto Z$  is point-wise sequentially continuous on uniformly bounded subsets of  $\mathcal{B}_b(Y)$  if for any uniformly bounded point-wise convergent sequence  $(g_n)_{n \in \mathbb{N}}$  in  $\mathcal{B}_b(Y)$  with limit  $g \in \mathcal{B}_b(Y)$  we have that  $\lim F(g_n) = F(g)$ .

Given a function space  $\mathcal{F} \subseteq \mathbb{R}^X$ . We say that a set  $S \subset \mathcal{F}$  is point-wise sequentially closed if  $S$  contains all the limits which are in  $\mathcal{F}$  of point-wise convergent sequences  $(g_n)_{n \in \mathbb{N}}$  whose elements  $g_n$  are in  $S$ . We say that a set  $\bar{S}$  is the point-wise sequential closure of a set  $S$  if  $\bar{S}$  is the smallest point-wise sequentially closed set containing  $S$ . A set  $S$  is point-wise sequentially dense in a set  $T$  if  $T$  is a subset of the sequential closure  $\bar{S}$  of  $S$ . (For more details on sequential continuity and measurable functions see [5] Section 3.2.)

**Proposition 1** Let  $K \subseteq Y$  be compact and let  $O \subset Y$  be open with compact closure such that  $K \subset O$ . Then there exists a continuous function  $f : Y \rightarrow [0, 1]$  such that  $f(K) = 1$  and  $f(Y \setminus O) = 0$ .

**Proof.** This follows by an application of the Urysohn Lemma (see [8] 15.6) to  $K$  and  $Y \setminus O \cup \{\infty\}$  in the one-point compactification (see [8] 19.2 and 19A)  $Y \cup \{\infty\}$  of  $Y$ , since any compact space is normal (see [8] 17.10).  $\square$

**Lemma 1** It holds that:

- The space  $\mathcal{B}(X)$  is point-wise sequentially closed in  $\mathbb{R}^X$ .
- The function-space  $\mathcal{B}_{\text{simp}}(Y)$  is point-wise sequentially dense in  $\mathcal{B}_b(Y)$ .
- The function-space  $\mathcal{C}_c(Y)$  is point-wise sequentially dense in  $\mathcal{B}_b(Y)$ .

**Proof.** (a) Is the well known fact that a limit of a point-wise convergent sequence of measurable functions is again measurable.

(b) Is a re-formulation of the fact that any measurable function is the point wise limit of a sequence of simple functions. (See for example Corollary 3.2.1 of [5].)

(c) Given an arbitrary compact set  $K$  we can by second countability and local compactness of  $Y$  choose a sequence  $(O_n)_{n \in \mathbb{N}}$  of open sets such that

$O_{n+1} \subset O_n$ ,  $\bigcap_n O_n = K$  and the closures  $\overline{O_n}$  are compact. By Proposition 1 we find continuous functions  $f_n$  such that  $f_n(K) = 1$  and  $f_n(Y \setminus O_n) = 0$ . Since  $\overline{O_n}$  is compact these functions  $f_n$  possess compact support. Thus,  $1_K = \lim_{n \in \mathbb{N}} f_n(x)$ , and  $1_K$  lies in the point-wise sequential closure of  $\mathcal{C}_c(Y)$ . Since any open set  $O$  is the countable union of compact sets, we see that also any function  $1_O$  and thus especially the function  $1_Y$  belongs to the sequential closure of  $\mathcal{C}_c(Y)$ . (That  $1_Y$  belongs to the sequential closure of  $\mathcal{C}_c(Y)$  can also be easily seen using a countable partition of unity.) Hence, any finite linear combination of function  $1_A$  with  $A \in \mathcal{Y}$  belongs to the sequential closure of  $\mathcal{C}_c(Y)$  and thus  $\mathcal{B}_b(Y)$  is a subset of the sequential closure of  $\mathcal{C}_c(Y)$ . So we obtain (c) from (b).  $\square$

**Lemma 2** *Any conditional integral  $I \in \mathcal{I}(X, Y)$  extends uniquely to a conditional integral  $\tilde{I} : X \times \mathcal{B}_b(Y) \mapsto \mathbb{R}$  such that for any  $x \in X$  the function  $\tilde{I}(x, \cdot)$  is point-wise sequentially continuous on uniformly bounded subsets of  $\mathcal{B}_b(Y)$ . Moreover, there exists a one-one correspondence between kernels and conditional integrals  $G : \mathcal{P}(X, Y) \rightarrow \mathcal{I}(X, Y)$  given by*

$$[G(P)](x, f) = \int f(y) P(x, dy) \text{ for all } f \in \mathcal{C}_c(Y), \quad (5)$$

or, if we prefer to consider the extensions  $\tilde{I}$  of the conditional integrals  $I$ , by

$$[\widetilde{G(P)}](x, g) = \int g(y) P(x, dy),$$

for all  $g \in \mathcal{B}_b(Y)$ .

We call the above extension  $\tilde{I}$  of a conditional integral  $I$  the *extended conditional integral*. By Lemma 1 there is a one-one correspondence between conditional integrals  $I$  and their extensions  $\tilde{I}$ .

**Proof of Lemma 2:** The proof consists of 3 steps. First we show that for a given conditional integral  $I \in \mathcal{I}(X, Y)$  there exists for any  $x \in X$  a unique measure  $P(x, \cdot)$  on  $(Y, \mathcal{Y})$ . Then we show that the integrals  $I(x, \cdot)$  on  $\mathcal{C}_c(Y)$  extend for arbitrary  $x \in X$  uniquely to extended integrals  $\tilde{I}(x, \cdot)$  on  $\mathcal{B}_b(Y)$ .

**Step 1:** Let  $I$  be a given conditional integral. According to the Riesz representation theorem, there exists for any  $x \in X$  a unique measure  $P(x, \cdot)$  on  $(Y, \mathcal{Y})$ , such that

$$I(x, f) = \int f(y) P(x, dy) \text{ for all } f \in \mathcal{C}_c(Y). \quad (6)$$

Thus, there exists for any  $x \in X$  a unique extended integral  $\tilde{I}(x, \cdot)$  such that

$$\tilde{I}(x, g) = \int g(y) P(x, dy) \quad \text{for all } g \in \mathcal{B}_b(Y). \quad (7)$$

Note that, by the dominated convergence theorem,  $\tilde{I}(x, \cdot)$  is sequentially point-wise continuous on uniformly bounded sets.  $\tilde{I}(x, \cdot)$  is also the unique extension of  $I(x, \cdot)$  from  $\mathcal{C}_c(Y)$  to  $\mathcal{B}_b(Y)$  which is sequentially point-wise continuous on uniformly bounded sets, since  $\{f \in \mathcal{C}_c(Y) \mid -1 \leq f \leq 1\}$  is point-wise sequentially dense in  $\{g \in \mathcal{B}_b(Y) \mid -1 \leq g \leq 1\}$  (The fact that  $\{f \in \mathcal{C}_c(Y) \mid -1 \leq f \leq 1\}$  is point-wise sequentially dense in  $\{g \in \mathcal{B}_b(Y) \mid -1 \leq g \leq 1\}$  is proved completely analogous as we proved (c) in Lemma 1.)

**Step 2:** In the second step we show that the functions  $x \mapsto \tilde{I}(x, g)$  are  $\mathcal{X}$ -measurable, for  $g \in \mathcal{B}_b(Y)$  arbitrary, i.e., we show that  $\tilde{I}$  is a conditional integral. Further we show that the unique corresponding function  $P : X \times \mathcal{Y}$ , defined in the first step, is a kernel.

Let  $\mathbb{R}^X$  be endowed with the topology of point-wise convergence. Define an operator  $T : \mathcal{B}_b(Y) \rightarrow \mathbb{R}^X$  by

$$[T(g)](x) = \tilde{I}(x, g).$$

The fact that, for arbitrary  $x \in X$ , the integral  $\tilde{I}(x, \cdot)$  is point-wise sequentially continuous on uniformly bounded sets of  $\mathcal{B}_b(Y)$  (where we take  $M = \mathcal{B}_b(Y)$  and  $V = \mathbb{R}$  in Definition 3) implies that  $T$  is also point-wise sequentially continuous (where we take  $M = \mathcal{B}_b(Y)$  and  $V = \mathbb{R}^X$  in Definition 3).

Further,  $f \in \mathcal{C}_c(Y)$  implies by definition of  $T$  and the fact that  $I \in \mathcal{I}(X, Y)$  that

$$T(f) = [x \mapsto I(x, f)] \in \mathcal{B}(X), \quad (8)$$

i.e., we have that  $T(\mathcal{C}_c(Y)) \subseteq \mathcal{B}(X)$ .

By (8) together with Lemma 1 (c) and the point-wise sequential continuity of  $T$ , we obtain that  $T(\mathcal{B}_b(Y)) \subseteq \mathcal{B}(X)$ . In other words, we obtain that  $g \in \mathcal{B}_b(Y)$  implies that  $x \mapsto \tilde{I}(x, g)$  is  $\mathcal{X}$ -measurable. The fact that  $x \mapsto \tilde{I}(x, g)$  is  $\mathcal{X}$ -measurable implies in the case that  $g$  is the characteristic function of a set  $B$  that  $x \mapsto P(x, B)$  is  $\mathcal{X}$ -measurable. Thus,  $P$  is a kernel and (as already noted in the first step) by the Riesz representation theorem unique.

In the first two steps we have shown that to an integral  $I \in \mathcal{I}(X, Y)$  there corresponds a unique kernel  $P \in \mathcal{P}(X, Y)$  and a unique extended integral  $\tilde{I}$ . Further we know by equation (6) and (5) that this correspondence is given by  $G^{-1}$ . In the third step we show that to any  $P \in \mathcal{P}(X, Y)$  there corresponds a unique  $I = G(P) \in \mathcal{I}(X, Y)$ .

**Step 3:** We show now that any kernel  $P$  corresponds to an unique integral  $I$ . That any kernel  $P$  gives us by formula (7) for any  $x$  an extended integral  $\tilde{I}(x, \cdot)$  is trivial. To show that  $\tilde{I}$  is a conditional extended integral note that for any simple function  $g = \sum_{i=1}^n \gamma_i \mathbf{1}_{B_i} \in \mathcal{B}_{simp}$  we have:

$$\tilde{I}(x, g) = \sum_i \gamma_i P(x, B_i).$$

So for  $g \in \mathcal{B}_{simp}$  the function  $x \mapsto \tilde{I}(x, g)$  is a finite sum of  $\mathcal{X}$ -measurable functions and thus itself  $\mathcal{X}$ -measurable. It remains to be shown that  $x \mapsto \tilde{I}(x, g)$  is for any  $g \in \mathcal{B}_b(Y)$  a  $\mathcal{X}$ -measurable function. We do this by arguments analogous to the arguments provided in step 2 as will be explained in the following.

Let  $T$  denote the operator defined in step 2. Recall that  $T$  is point-wise sequentially continuous. Furthermore,  $f \in \mathcal{B}_{simp}(Y)$  implies (by definition of  $T$  and the fact that for  $g \in \mathcal{B}_{simp}(Y)$  the function  $x \mapsto \tilde{I}(x, g)$  is  $\mathcal{X}$ -measurable) that:

$$T(f) = [x \rightarrow \tilde{I}(x, f)] \in \mathcal{B}(X), \quad (9)$$

i.e., we have that  $T(\mathcal{B}_{simp}(Y)) \subseteq \mathcal{B}(X)$ .

By (9) together with Lemma 1 (b) and point-wise sequential continuity of  $T$ , we obtain that  $T(\mathcal{B}_b(Y)) = \mathcal{B}(X)$ . In other words, we obtain that  $g \in \mathcal{B}$  implies that  $x \mapsto \tilde{I}(x, g)$  is  $\mathcal{X}$ -measurable.  $\square$

Now we define weak differentiability of conditional integrals and kernels.

**Definition 4** Let  $\Theta$  be an open interval in  $\mathbb{R}$  and let  $\vartheta \mapsto I_\vartheta$  be a path in (mapping from  $\Theta$  to) the space  $\mathcal{I}(X, Y)$ . We say that  $\vartheta \mapsto I_\vartheta$  is weakly differentiable if

$$\frac{dI_\vartheta(x, f)}{d\vartheta} \text{ exists for all } (x, f) \in X \times \mathcal{C}_c(Y)$$

If  $\vartheta \mapsto I_\vartheta$  is weakly differentiable then we say that it is bounded weakly differentiable if

$$\sup_{\substack{f \in \mathcal{C}_c(Y) \\ |f| \leq 1}} \left| \frac{dI_\vartheta(x, f)}{d\vartheta} \right| < \infty,$$

for any  $x \in X$ .

We say that a path  $\theta \mapsto P_\theta$  in the space  $\mathcal{P}(X, Y)$  of kernels is bounded differentiable if the corresponding path  $\theta \mapsto G(P_\theta)$  in the space  $\mathcal{I}(X, Y)$  of conditional integrals is bounded weakly differentiable.



**Theorem 1** *If the path  $\vartheta \mapsto P_\vartheta$  in the space  $\mathcal{P}(X, Y)$  is bounded weakly differentiable, then the weak derivative can be represented by a path  $\vartheta \mapsto P'_\vartheta$  in the space  $\mathcal{P}(X, Y)$ . The connection between  $\vartheta \mapsto P_\vartheta$  and  $\vartheta \mapsto P'_\vartheta$  is given by*

$$\int f(y)P'_\vartheta(x, dy) = \frac{d \int f(y)P_\vartheta(x, dy)}{d\vartheta}.$$

**Proof.** Let  $I_\vartheta = G(P_\vartheta)$  be the corresponding path in the space of conditional integrals. Define for any  $(x, f) \in X \times \mathcal{C}_c(Y)$  the function  $I'_\vartheta(x, f)$  by

$$I'_\vartheta(x, f) = \frac{dI_\vartheta(x, f)}{d\vartheta}$$

Let  $(h_n)_{n \in \mathbb{N}}$  be an arbitrary sequence of positive reals which goes to 0. Then for  $f \in \mathcal{C}_c$  we have:

$$x \mapsto I'_\vartheta(x, f) = x \mapsto \frac{dI_\vartheta(x, f)}{d\vartheta} = x \mapsto \lim_{n \rightarrow \infty} \frac{I_{\vartheta+h_n}(x, f) - I_\vartheta(x, f)}{h_n}.$$

Thus,  $x \mapsto I'_\vartheta(x, f)$  is for  $f \in \mathcal{C}_c(Y)$  a limit of a sequence of  $\mathcal{X}$ -measurable functions and therefore itself  $\mathcal{X}$ -measurable. Furthermore,  $I'(x, \cdot)$  is by the condition of boundedness in the definition of bounded weakly differentiable for any  $x \in X$  norm-bounded; i.e.,  $I'(x, \cdot)$  is bounded on uniformly bounded subsets of  $\mathcal{C}_c(Y)$ . Thus,  $I'(x, \cdot)$  is for any  $x \in X$  an integral and  $I'(\cdot, \cdot)$  is thus itself a conditional integral. By the correspondence between conditional integrals and kernels we obtain a kernel  $P' = G^{-1}(I')$ . The formula connecting  $P'$  and  $P$  is clear from the correspondence between  $P', P$  and  $I', I$  and the definition of  $I'$ .  $\square$

### 3 Jordan Decomposition of Weak Derivatives of Markov Kernels

**Definition 5** *Given a kernel  $P \in \mathcal{P}(X, Y)$  we define the absolute value  $|P|$  of the kernel as follows:*

$$|P|(x, B) = \sup_{\substack{A \in \mathcal{Y} \\ A \subseteq B}} 2 \cdot P(x, A) - P(x, B), \quad x \in X, B \in \mathcal{Y}.$$

**Lemma 3** *The absolute value  $|P|$  of a kernel  $P \in \mathcal{P}(X, Y)$  is again a kernel.*

**Proof:** That the absolute value  $|P|(x, \cdot)$  is a finite measure is a well known fact and it remains to be shown that the function

$$x \mapsto |P|(x, B)$$

is  $\mathcal{X}$ -measurable.

Let  $\mathcal{A}$  be the set-field generated by a countable basis of the topology of  $Y$ . Then,  $\mathcal{A}$  is countable and generates the  $\sigma$ -field  $\mathcal{Y}$ . For any set  $B \in \mathcal{Y}$  and any measure  $\mu$  on  $(Y, \mathcal{Y})$  there exists a sequence  $(A_n)_{n \in \mathbb{N}}$  of sets  $A_n \in \mathcal{A}$  such that  $\lim \mu(A_n \triangle B) = 0$  (see [6] Lemma A.24). Thus, the function

$$x \mapsto |P|(x, B)$$

is the point-wise supremum over the countable family

$$\left\{ x \mapsto 2 \cdot P(x, A) - P(x, B) : A \in \mathcal{A} \text{ and } A \subseteq B \right\}$$

of  $\mathcal{X}$ -measurable functions and thus itself a  $\mathcal{X}$ -measurable function on  $X$ .  $\square$

**Definition 6** We say that a kernel is positive if  $P(x, B) \geq 0$  for all  $(x, B) \in X \times \mathcal{Y}$ . We say that a pair of kernels  $(P^+, P^-)$  forms a decomposition of a kernel  $P$  if  $P^+$  and  $P^-$  are positive kernels and  $P(x, B) = P^+(x, B) - P^-(x, B)$ . We say that this decomposition is minimal or Jordan if for any other decomposition  $(Q^+, Q^-)$  of  $P$  we have  $P^+(x, B) \leq Q^+(x, B)$  and  $P^-(x, B) \leq Q^-(x, B)$ .

**Corollary 1** Any kernel  $P \in \mathcal{P}(X, Y)$  possesses a Jordan decomposition.

**Proof:** For  $(x, B) \in X \times \mathcal{Y}$  define

$$P^+(x, B) := \frac{|P|(x, B) + P(x, B)}{2}$$

and

$$P^-(x, B) := \frac{|P|(x, B) - P(x, B)}{2}.$$

Then,  $P^+(x, B), P^-(x, B) \geq 0$  and  $P^+(x, \cdot), P^-(x, \cdot)$  are measures, and  $x \mapsto P^+(x, B)$  as well as  $x \mapsto P^-(x, B)$  are  $\mathcal{X}$ -measurable functions on  $X$ . It is also clear that the decomposition is minimal.  $\square$

**Theorem 2** Suppose that the path  $\vartheta \mapsto P_\vartheta$  in the space  $\mathcal{P}(X, Y)$  is bounded weakly differentiable and that for any  $\theta$  the kernel  $P_\theta$  is Markov. Then there exist for any  $\vartheta$  Markov kernels  $Q_\vartheta^+$  and  $Q_\vartheta^-$  from  $X$  to  $Y$  and a  $\mathcal{X}$ -measurable function  $c_\vartheta : X \rightarrow \mathbb{R}$  such that the weak derivative  $P'_\vartheta$  of  $P_\vartheta$  decomposes in the form

$$P'_\vartheta(x, B) = c_\vartheta(x) (Q_\vartheta^+(x, B) - Q_\vartheta^-(x, B)) \quad \forall (x, B) \in X \times \mathcal{Y}.$$

**Proof:** By Theorem 1, the weak derivative  $P'_\vartheta$  is for any  $\vartheta$  a kernel and by the Corollary 1,  $P'_\vartheta$  possesses a Jordan decomposition  $(P_\vartheta^+, P_\vartheta^-)$ , i.e.,  $P'_\vartheta = P_\vartheta^+ - P_\vartheta^-$  and  $P_\vartheta^+, P_\vartheta^-$  are positive kernels. Since the  $P_\vartheta$  are Markov kernels we have  $P_\vartheta^+(x, Y) = P_\vartheta^-(x, Y)$ . Let  $c_\vartheta : X \rightarrow \mathbb{R}$  be defined by

$$c_\vartheta(x) := P_\vartheta^+(x, Y) = P_\vartheta^-(x, Y).$$

Since  $P_\vartheta^+$  is a kernel, the function  $c(\cdot)$  is  $\mathcal{X}$ -measurable. Let

$$Q_\vartheta^+(x, B) := \frac{1}{c(x)} P_\vartheta^+(x, B) \text{ for all } x \text{ with } c(x) > 0,$$

$$Q_\vartheta^-(x, B) := \frac{1}{c(x)} P_\vartheta^-(x, B) \text{ for all } x \text{ with } c(x) > 0$$

and let for an arbitrary fixed probability measure  $\mu$ , arbitrary  $x$  with  $c_\vartheta(x) = 0$  and arbitrary  $B \in \mathcal{Y}$

$$Q_\vartheta^+(x, B) = Q_\vartheta^-(x, B) = \mu(B).$$

Then  $Q_\vartheta^+$  as well as  $Q_\vartheta^-$  are Markov kernels. □

**Remark 3** *This specific decomposition  $(c_\vartheta(\cdot), Q_\vartheta^+, Q_\vartheta^-)$  is only possible because the kernels  $P'_\vartheta$  stem from weak differentiation of a Markov kernel valued function  $\theta \mapsto P_\vartheta$ .*

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