

## A SIMPLE PROOF OF DUALITY THEOREM FOR MONGE-KANTOROVICH PROBLEM

TOSHIO MIKAMI

*To the memory of the late Professor Nobuyuki Suita*

### Abstract

We give a simple proof of the duality theorem for the Monge-Kantorovich problem in the Euclidean setting. The selection lemma which is useful in the theory of stochastic optimal controls plays a crucial role.

### 1. Introduction

Let  $P_0$  and  $P_1$  be Borel probability measures on  $\mathbf{R}^d$  and  $\mathcal{A}(P_0, P_1)$  denote the set of all  $\mu \in \mathcal{M}_1(\mathbf{R}^d \times \mathbf{R}^d)$  for which  $\mu(dx \times \mathbf{R}^d) = P_0(dx)$  and  $\mu(\mathbf{R}^d \times dx) = P_1(dx)$ , where  $\mathcal{M}_1(\mathbf{R}^d \times \mathbf{R}^d)$  denotes the complete separable metric space, with a weak topology, of Borel probability measures on  $\mathbf{R}^d \times \mathbf{R}^d$  (see e.g. [1]). Take also a Borel measurable  $c(\cdot, \cdot) : \mathbf{R}^d \times \mathbf{R}^d \mapsto [0, \infty)$ .

The study of a minimizer of the following  $\mathcal{T}(P_0, P_1)$  is called the Monge-Kantorovich problem which has been studied by many authors and which has been applied to many fields (see [2, 4, 6, 9, 10] and the references therein):

$$(1.1) \quad \mathcal{T}(P_0, P_1) := \inf \left\{ \int_{\mathbf{R}^d \times \mathbf{R}^d} c(x, y) \mu(dx dy) \mid \mu \in \mathcal{A}(P_0, P_1) \right\}.$$

The duality theorem for  $\mathcal{T}(P_0, P_1)$  plays a crucial role in the proof of the Monge-Kantorovich problem and has been proved for a wide class of functions  $c(\cdot, \cdot)$  (see [5, 8–10]).

They say that the duality theorem for  $\mathcal{T}(P_0, P_1)$  holds if

$$(1.2) \quad \mathcal{T}(P_0, P_1) = \sup \left\{ \int_{\mathbf{R}^d} \psi(1, y) P_1(dy) - \int_{\mathbf{R}^d} \psi(0, x) P_0(dx) \right\},$$

where the supremum is taken over all  $\psi(t, \cdot) \in L^1(P_t)$  ( $t = 0, 1$ ) for which

---

Partially supported by the Grant-in-Aid for Scientific Research, No. 15340047, 15340051 and 16654031, JSPS.

Received December 14, 2004; revised June 10, 2005.

$$(1.3) \quad \psi(1, y) - \psi(0, x) \leq c(x, y).$$

In [7] we obtained a stochastic control version of (1.2)–(1.3) and gave a new approach to  $h$ -path processes for diffusion processes as its application. We also showed that its zero noise limit yields (1.2)–(1.3) (see [8]).

In this paper we give a simple proof for (1.2)–(1.3) since known proofs for (1.2)–(1.3) are complicated. Indeed, Kellerer used a functional version of Choquet's capacitability theorem (see [5] and also [9]) and Villani did a minimax principle (see [10, sect. 1.1]).

Our proof relies on the Legendre duality of a lower semicontinuous convex function of Borel probability measures on  $\mathbf{R}^d$  and on the selection lemma which is useful in the theory of stochastic optimal controls (see the proof of Theorem 2.1).

I would like to dedicate this paper to the late Professor Nobuyuki Suita who used to be very nice to me.

## 2. A simple proof for Duality Theorem

We first describe our assumption.

(A.1)  $c \in C(\mathbf{R}^d \times \mathbf{R}^d : [0, \infty))$  and  $c(x, y) \rightarrow \infty$  as  $|y - x| \rightarrow \infty$ , and  $\inf\{c(x, y) \mid y \in \mathbf{R}^d\}$  is bounded.

(A.2)  $\mathcal{F}(P_0, P_1)$  is finite.

(A.3)  $P_0$  is absolutely continuous with respect to the Lebesgue measure  $dx$ .

We give a simple proof to the following which can be obtained from the known result (see [5] and also [9, 10]).

**THEOREM 2.1 (Duality Theorem).** *Suppose that (A.1)–(A.3) hold. Then (1.2)–(1.3) holds.*

*Proof.* We prove (1.2), where the supremum is taken over all  $\psi(t, \cdot) \in C_b(\mathbf{R}^d)$  ( $t = 0, 1$ ) for which (1.3) holds. This implies the duality theorem for  $\mathcal{F}(P_0, P_1)$  since (1.2)–(1.3) with “=” replaced by “ $\geq$ ” holds and since  $C_b(\mathbf{R}^d) \subset L^1(P_t)$  ( $t = 0, 1$ ).

The proof is divided into the following (2.1)–(2.3):

$$(2.1) \quad \mathcal{F}(P_0, P_1) = \sup_{f \in C_b(\mathbf{R}^d)} \left\{ \int_{\mathbf{R}^d} f(y) P_1(dy) - \mathcal{F}_{P_0}^*(f) \right\},$$

where for  $f \in C_b(\mathbf{R}^d)$ ,

$$\mathcal{F}_{P_0}^*(f) := \sup_{P \in \mathcal{M}_1(\mathbf{R}^d)} \left\{ \int_{\mathbf{R}^d} f(y) P(dy) - \mathcal{F}(P_0, P) \right\}.$$

For  $f \in C_b(\mathbf{R}^d)$ ,

$$(2.2) \quad \varphi(x; f) := \sup_{y \in \mathbf{R}^d} \{f(y) - c(x, y)\} \in C_b(\mathbf{R}^d),$$

$$(2.3) \quad \mathcal{F}_{P_0}^*(f) = \int_{\mathbf{R}^d} \varphi(x; f) P_0(dx).$$

We first prove (2.1). We only have to prove  $\mathcal{F}(P_0, \cdot) : \mathcal{M}_1(\mathbf{R}^d) \mapsto [0, \infty]$  is lower semicontinuous and convex. Indeed, this and (A.2) implies (2.1) from [1, Theorem 2.2.15 and Lemma 3.2.3], by putting  $\mathcal{F}(P_0, P) = \infty$  for  $P \notin \mathcal{M}_1(\mathbf{R}^d)$ .

Suppose that  $Q_n \rightarrow Q$  weakly as  $n \rightarrow \infty$ . Then it is easy to see that  $\bigcup_{n \geq 1} \mathcal{A}(P_0, Q_n)$  is tight in  $\mathcal{M}_1(\mathbf{R}^d)$ . Take  $\mu_n \in \mathcal{A}(P_0, Q_n)$  ( $n \geq 1$ ) for which

$$(2.4) \quad \mathcal{F}(P_0, Q_n) + \frac{1}{n} > \int_{\mathbf{R}^d \times \mathbf{R}^d} c(x, y) \mu_n(dxdy) \geq \mathcal{F}(P_0, Q_n).$$

For any convergent subsequence  $\{\mu_{k(n)}\}_{n \geq 1}$  of  $\{\mu_n\}_{n \geq 1}$  and its weak limit  $\mu_0$ ,  $\mu_0 \in \mathcal{A}(P_0, Q)$  and

$$(2.5) \quad \liminf_{n \rightarrow \infty} \int_{\mathbf{R}^d \times \mathbf{R}^d} c(x, y) \mu_{k(n)}(dxdy) \geq \int_{\mathbf{R}^d \times \mathbf{R}^d} c(x, y) \mu_0(dxdy)$$

since  $c \geq 0$  from (A.1). Hence  $\mathcal{F}(P_0, \cdot) : \mathcal{M}_1(\mathbf{R}^d) \mapsto [0, \infty]$  is lower semicontinuous.  $\mathcal{F}(P_0, \cdot) : \mathcal{M}_1(\mathbf{R}^d) \mapsto [0, \infty]$  is also convex since for any  $P, Q \in \mathcal{M}_1(\mathbf{R}^d)$  and  $\lambda \in (0, 1)$ ,

$$\{\lambda\mu + (1-\lambda)\nu \mid \mu \in \mathcal{A}(P_0, P), \nu \in \mathcal{A}(P_0, Q)\} \subset \mathcal{A}(P_0, \lambda P + (1-\lambda)Q).$$

We next prove (2.2). From (A.1), for  $f \in C_b(\mathbf{R}^d)$  and  $x \in \mathbf{R}^d$ ,

$$(2.6) \quad -\infty < \inf_{y \in \mathbf{R}^d} f(y) - \sup_{x \in \mathbf{R}^d} \left\{ \inf_{y \in \mathbf{R}^d} c(x, y) \right\} \leq \varphi(x; f) \leq \sup_{y \in \mathbf{R}^d} f(y) < \infty,$$

which implies that  $\varphi(\cdot; f)$  is bounded.

From (A.1), the following set is not empty for any  $x \in \mathbf{R}^d$  and is bounded on every bounded subset of  $\mathbf{R}^d$ :

$$(2.7) \quad D_x := \{y \in \mathbf{R}^d \mid \varphi(x; f) = f(y) - c(x, y)\}.$$

Suppose that  $x_n \rightarrow x$  as  $n \rightarrow \infty$ . Take  $y_n \in D_{x_n}$  and  $y \in D_x$ . Then there exist a convergent subsequence  $\{y_{k(n)}\}_{n \geq 1}$  and  $\tilde{y}$  such that  $y_{k(n)} \rightarrow \tilde{y}$  as  $n \rightarrow \infty$  and such that

$$(2.8) \quad \begin{aligned} \limsup_{n \rightarrow \infty} \varphi(x_n; f) &= \lim_{n \rightarrow \infty} \{f(y_{k(n)}) - c(x_{k(n)}, y_{k(n)})\} \\ &= f(\tilde{y}) - c(x, \tilde{y}) \leq \varphi(x; f). \end{aligned}$$

The following together with (2.8) implies that  $\varphi(\cdot; f) \in C(\mathbf{R}^d)$ :

$$(2.9) \quad \liminf_{n \rightarrow \infty} \varphi(x_n; f) \geq \lim_{n \rightarrow \infty} \{f(y) - c(x_n, y)\} = \varphi(x; f).$$

We prove (2.3) to complete the proof. For  $f \in C_b(\mathbf{R}^d)$ ,

$$(2.10) \quad \mathcal{T}_{P_0}^*(f) = \sup_{P \in \mathcal{M}_1(\mathbf{R}^d)} \left\{ \sup \left\{ \int_{\mathbf{R}^d \times \mathbf{R}^d} (f(y) - c(x, y)) \mu(dx dy) \mid \mu \in \mathcal{A}(P_0, P) \right\} \right\} \\ \leq \int_{\mathbf{R}^d} \varphi(x; f) P_0(dx).$$

(A.1) implies that the set  $\bigcup_{|x| \leq r} (\{x\} \times D_x)$  is compact for any  $r > 0$ . Indeed, the set  $\bigcup_{|x| \leq r} (\{x\} \times D_x)$  is bounded as we mentioned in (2.7) and is closed since the set  $D := \{(x, y) \in \mathbf{R}^d \times \mathbf{R}^d \mid \varphi(x; f) = f(y) - c(x, y)\}$  is closed from (2.2) and since

$$\bigcup_{|x| \leq r} (\{x\} \times D_x) = D \cap \{(x, y) \in \mathbf{R}^d \times \mathbf{R}^d \mid |x| \leq r\}.$$

Hence there exists a measurable function  $u : \mathbf{R}^d \mapsto \mathbf{R}^d$  such that  $u(x) \in D_x$ ,  $dx$ -a.e. by the selection lemma (see [3, p. 199]). In particular, from (A.3) and (2.10),

$$(2.11) \quad \mathcal{T}_{P_0}^*(f) \leq \int_{\mathbf{R}^d \times \mathbf{R}^d} \{f(y) - c(x, y)\} P_0(dx) \delta_{u(x)}(dy) \leq \mathcal{T}_{P_0}^*(f). \quad \text{Q.E.D.}$$

#### REFERENCES

- [1] J. D. DEUSCHEL AND D. W. STROOCK, Large deviations, Pure and applied mathematics **137**, Academic Press, Inc., Boston, MA, 1989.
- [2] L. C. EVANS, Partial differential equations and Monge-Kantorovich mass transfer, Current developments in mathematics, 1997, Int. Press and Cambridge, Boston, MA, 1999, 65–126.
- [3] W. H. FLEMING AND R. W. RISHEL, Deterministic and stochastic optimal control, Springer-Verlag, Berlin, Heidelberg, New York, Tokyo, 1975.
- [4] W. GANGBO AND R. J. MCCANN, The geometry of optimal transportation, Acta Math. **177** (1996), 113–161.
- [5] H. G. KELLERER, Duality theorems for marginal problems, Z. Wahrsch. Verw. Gebiete **67** (1984), 399–432.
- [6] T. MIKAMI, Monge’s problem with a quadratic cost by the zero-noise limit of  $h$ -path processes, Probab. Theory Related Fields **129** (2004), 245–260.
- [7] T. MIKAMI AND M. THIEULLEN, Duality theorem for stochastic optimal control problem, Hokkaido University preprint series **652**, 2004.
- [8] T. MIKAMI AND M. THIEULLEN, Optimal transportation problem by stochastic optimal control, Hokkaido University preprint series **690**, 2005.
- [9] S. T. RACHEV AND L. RÜSCHENDORF, Mass transportation problems, I: Theory, II: Application, Springer-Verlag, Berlin, Heidelberg, New York, Tokyo, 1998.
- [10] C. VILLANI, Topics in optimal transportation, Graduate studies in mathematics **58**, Amer. Math. Soc., Providence, RI, 2003.

DEPARTMENT OF MATHEMATICS  
HOKKAIDO UNIVERSITY  
SAPPORO 060-0810  
JAPAN  
E-mail: mikami@math.sci.hokudai.ac.jp