## The definition of optimal transport paths

Let $X$ be a compact convex subset of a Euclidean space $\mathbb{R}^{m}$. Suppose

$$
\begin{equation*}
\mathbf{a}=\sum_{i=1}^{k} a_{i} \delta_{x_{i}} \text { and } \mathbf{b}=\sum_{j=1}^{l} b_{j} \delta_{y_{j}} \tag{0.1}
\end{equation*}
$$

are two atomic probability measures on $X$. A transport path from $a$ to $b$ is a weighted directed graph $\mathbf{G}$ consists of a vertex set $V(\mathbf{G})$, a directed edge set $E(\mathbf{G})$ and a weight function

$$
w: E(\mathbf{G}) \rightarrow(0,+\infty)
$$

such that $\left\{x_{1}, x_{2}, \cdots, x_{k}\right\} \cup\left\{y_{1}, y_{2}, \cdots, y_{l}\right\} \subset V(\mathbf{G})$ and for any vertex $v \in V(\mathbf{G})$,

$$
\sum_{\substack{e \in E(\mathbf{G})  \tag{0.2}\\
e^{-}=v}} w(e)=\sum_{\substack{e \in E(\mathbf{G}) \\
e^{+}=v}} w(e)+\left\{\begin{array}{cc}
a_{i}, & \text { if } v=x_{i} \text { for some } i=1, \cdots, k \\
-b_{j}, & \text { if } v=y_{j} \text { for some } j=1, \cdots, l \\
0, & \text { otherwise }
\end{array}\right.
$$

where $e^{-}$and $e^{+}$denotes the starting and ending endpoints of each edge $e \in E(\mathbf{G})$.
The balance equation (0.2) simply means that the total mass flows into $v$ equals to the total mass flows out of $v$. When $\mathbf{G}$ is viewed as a polyhedral chain or current, (0.2) can be simply expressed as $\partial \mathbf{G}=\mathbf{b}-\mathbf{a}$. Also, when $\mathbf{G}$ is viewed as a vector valued measure, the balance equation is simply $\operatorname{div}(\mathbf{G})=\mathbf{a}-\mathbf{b}$ in the sense of distributions.

Let $\alpha \leq 1$ be a parameter. The $\mathbf{M}_{\alpha}$ cost function on a transport path $\mathbf{G}$ is defined by

$$
\mathbf{M}_{\alpha}(\mathbf{G}) \equiv \sum_{e \in E(\mathbf{G})}[w(e)]^{\alpha} \text { length }(e)
$$

for any transport path $\mathbf{G}$ from $\mathbf{a}$ to $\mathbf{b}$. Any $\mathbf{M}_{\alpha}$ minimizer in the family of all transport paths from $\mathbf{a}$ to $\mathbf{b}$ is called an optimal transport path.

Now, we can talk about transport paths between general probability measures. Let $\mu^{+}, \mu^{-}$be any two probability measures on $X$. Extending the above definition, we say a vector measure $\mathbf{T}$ on $X$ is a transport path from $\mu^{+}$to $\mu^{-}$if there exist two sequences $\left\{\mathbf{a}_{i}\right\},\left\{\mathbf{b}_{i}\right\}$ of atomic probability measures on $X$ with a corresponding sequence of transport paths $\mathbf{G}_{i}$ from $\mathbf{a}_{i}$ to $\mathbf{b}_{i}$ such that

$$
\mathbf{a}_{i} \rightharpoonup \mu^{+}, \mathbf{b}_{i} \rightharpoonup \mu^{-}, \mathbf{G}_{i} \rightharpoonup \mathbf{T}
$$

weakly as probability measures and vector measures. Note that for any such $\mathbf{T}$, $\operatorname{div}(\mathbf{T})=\mu^{+}-\mu^{-}$in the sense of distributions. Also, given any $\alpha \in[0,1]$, for any transport path $\mathbf{T}$ from $\mu^{+}$to $\mu^{-}$, we define its $\mathbf{M}_{\alpha}$ cost to be

$$
\mathbf{M}_{\alpha}(\mathbf{T}):=\inf \lim \inf _{i \rightarrow \infty} \mathbf{M}_{\alpha}\left(\mathbf{G}_{i}\right)
$$

where the infimum is over the set of all possible approximating graph sequence $\left\{\mathbf{a}_{i}, \mathbf{b}_{i}, \mathbf{G}_{i}\right\}$ of $\mathbf{T}$.

