## Research Article

# Invariants for Weighted Digraphs under One-Sided State Splittings 

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Using Matrix-Forest theorem and Matrix-Tree theorem, we present some invariants for weighted digraphs under state in-splittings or out-splittings.

## 1. Introduction

State in-splittings and out-splittings are very important operations in the theory of one-sided, or two-sided Markov shifts ( $[1,2]$ ). Lind and Tuncel introduced a spanning tree invariant for Markov shifts in [3]. Spanning tree invariants are further studied in [4-6]. Motivated by these works, we consider some other graph structures like cycles and forests and present some invariants for weighted digraphs under state insplittings or out-splittings.

Firstly we give some basic definitions in graph theory and a brief introduction of Matrix-Forest theorem for digraphs. Readers can refer to [7,8] for more details.

In this paper, a digraph is an ordered pair $D=(V, E)$ of finite sets, where $V$ is called the vertex set and $E \subseteq V \times V$ is called the edge set. For an edge $(u, v) \in E, u$ and $v$ are called the initial and terminal ends of the edge, respectively. The number of edges having $u$ as the initial end is defined to be the outdegree of $u$ and denoted by $d(u)$. The number of edges having $v$ as the terminal end is defined to be the indegree of $v$. A walk of length $n$ is a sequence of edges $\left\{\left(u_{i}, u_{i+1}\right)\right\}(i=$ $1, \ldots, n)$ and can be denoted by $\left(u_{1}, u_{2}, \ldots, u_{n+1}\right)$; moreover, if $u_{n+1}$ is the same as $u_{1}$, we call the walk a closed one. A directed forest is a digraph without closed walks such that the indegree of each vertex is no more than one. The vertices with indegree zero of a forest are called roots. We say that $D_{0}=\left(V_{0}, E_{0}\right)$ is a spanning subgraph of $D$ if $V_{0}=V$ and $E_{0} \subseteq E$.

Suppose that $D$ is a digraph with vertex set $V(D)=\{1$, $\ldots, n\}$. Let $w: E(D) \rightarrow \mathbb{R}^{+}$be a weight function on the edge
set. We then say that $\mathscr{D}=(D, w)$ is a weighted digraph and $M=(w(i, j))_{n \times n}$ is the weight matrix of $\mathscr{D}$. The Kirchhoff matrix of $\mathscr{D}$ is defined as $L=R-M$, where $R=\left(r_{i, j}\right)$ is a diagonal matrix and $r_{i, i}=\sum_{j=1}^{n} w(i, j)$. The product of the weights of all edges that belong to a subgraph $\mathscr{H}$ of $\mathscr{D}$ is defined to be the weight of $\mathscr{H}$ and denoted by $w(\mathscr{H})$.

Let $\mathscr{F}(\mathscr{D})=\mathscr{F}$ be the set of all spanning rooted forests of $\mathscr{D}$ and $\mathscr{F}^{i \rightarrow j}(\mathscr{D})=\mathscr{F}^{i \rightarrow j}$ the set of those spanning rooted forests of $\mathscr{D}$ such that $i$ and $j$ belong to the same tree rooted at $i$. For a matrix $A, A^{i, j}$ denotes the cofactor of the $(i, j)$-entry of $A$. The Matrix-Forest theorem then states as follows.

Lemma 1 (cf. [8]). Let $\mathscr{D}=(D, w)$ be a weighted digraph. Let $L$ be the Kirchhoff matrix of $\mathscr{D}$. Then one has
(1) $\sum_{F \in \mathscr{F}} w(F)=\operatorname{det}(I+L)$;
(2) for any $i, j \in V(D), \sum_{F \in \mathscr{F}^{i \rightarrow j}} w(F)=(I+L)^{i, j}$.

## 2. Invariants for Weighted Digraphs under State In-Splitting

Before giving the main result, we recall the definition of state in-splitting.

Definition 2. Let $\mathscr{D}=(D, w)$ be a weighted digraph. For a vertex $u$ of $D, E^{u}$ denotes the set of edges of $D$ with terminal end $u$. The state in-splitting of $\mathscr{D}$ at $u$ induces a new weighted digraph $\widetilde{\mathscr{D}}=(\widetilde{D}, \widetilde{w})$ in the following way: let $\mathcal{\mathcal { S }}=$
$\left\{S_{1}, S_{2}, \ldots, S_{r}\right\}$ be a partition of $E^{u}$. The vertex set of the new digraph is $V(\widetilde{D})=(V(D) \backslash\{u\}) \bigcup\left\{u_{1}, u_{2}, \ldots, u_{r}\right\}$. The edge set $E(\widetilde{D})$ and weight $\widetilde{w}$ of $\widetilde{\mathscr{D}}$ are defined as follows.
(i) For $x, y \in V(D) \backslash\{u\},(x, y) \in E(\widetilde{D})$ if and only if $(x, y) \in E(D)$ and in this case $\widetilde{w}(x, y)=w(x, y)$.
(ii) For $x \in V(D) \backslash\{u\},\left(x, u_{i}\right) \in E(\widetilde{D})$ if and only if $(x, u) \in$ $S_{i}$ and in this case $\widetilde{w}\left(x, u_{i}\right)=w(x, u)$.
(iii) For $x \in V(D) \backslash\{u\},\left(u_{i}, x\right) \in E(\widetilde{D})$ if and only if $(u, x) \in$ $E(D)$ and in this case $\widetilde{w}\left(u_{i}, x\right)=w(u, x)$.
(iv) If $(u, u) \in S_{i}$, then $\left(u_{j}, u_{i}\right) \in E(\widetilde{D})$, for $j=1,2, \ldots, r$, and in this case $\widetilde{w}\left(u_{j}, u_{i}\right)=w(u, u)$.
For more details about state splittings, readers can refer to $[2,3,9]$. Now we give the definition of our new invariant.

Definition 3. Let $\mathscr{D}=(D, w)$ be a weighted digraph. We define $W_{k}(\mathscr{D})(k \geq 1)$ as

$$
\begin{equation*}
W_{k}(\mathscr{D})=\sum_{v} d(v) \sum_{C \in C_{v}^{k}} w(C), \tag{1}
\end{equation*}
$$

where $v$ runs over $V(D)$ and $C_{v}^{k}$ denotes the set of closed walks of $\mathscr{D}$ with length $k$ at vertex $v$. Furthermore, we define the generating function $W_{\mathscr{D}}(t)$ as

$$
\begin{equation*}
W_{\mathscr{D}}(t)=\sum_{k \geq 1} W_{k}(\mathscr{D}) t^{k} \tag{2}
\end{equation*}
$$

Let $A$ be a square matrix. The trace of $A$ is defined to be the sum of the elements on the main diagonal and denoted by $\operatorname{tr}(A)$. For a digraph $D$, the diagonal matrix $O(D)=\left(o_{i, i}\right)$ denotes the outdegree matrix of $D$ that is, $o_{i, i}=d\left(v_{i}\right)$. Then we have the following result.

Theorem 4. Let $\mathscr{D}$ be a weighted digraph with weight matrix $M$. Then $W_{\mathscr{D}}(t)$ is an invariant under state in-splitting and can be computed in the following way:

$$
\begin{equation*}
W_{\mathscr{D}}(t)=\frac{\operatorname{tr}(O \cdot \operatorname{adj}(I-t M))}{\operatorname{det}(I-t M)}-\operatorname{tr}(O) \tag{3}
\end{equation*}
$$

Proof. We firstly prove the invariance of $W_{k}(\mathscr{D})$ for $k \geq 1$. Without loss of generality, if there is a loop at vertex $u$, we assume that it belongs to $S_{1}$, where $\mathcal{S}=\left\{S_{1}, S_{2}, \ldots, S_{r}\right\}$ denotes the partition of $E^{u}$ as in the definition of state insplitting.

We define the mapping

$$
\begin{equation*}
\varphi: \bigcup_{v \in V(D)} C_{v}^{k}(\mathscr{D}) \longrightarrow \bigcup_{v \in V(\widetilde{D})} C_{v}^{k}(\widetilde{\mathscr{D}}) \tag{4}
\end{equation*}
$$

in the following way: for a closed walk $C$ of $\mathscr{D}$ with length $k$, if $C=(u, u, \ldots, u)$, then $\varphi(C)=\left(u_{1}, u_{1}, \ldots, u_{1}\right)$; otherwise, we replace each maximum path of $C$ of the form $(v, u, u, \ldots, u)(v \neq u)$ with $\left(v, u_{i}, u_{1}, \ldots, u_{1}\right)$ if $(v, u) \in S_{i}$. it is not difficult to see that

$$
\begin{equation*}
\varphi: C_{v}^{k}(\mathscr{D}) \longrightarrow C_{v}^{k}(\widetilde{\mathscr{D}}) \tag{5}
\end{equation*}
$$



Figure 1
where $v \neq u$, and

$$
\begin{equation*}
\varphi: C_{u}^{k}(\mathscr{D}) \longrightarrow \bigcup_{i=1}^{r} C_{u_{i}}^{k}(\widetilde{\mathscr{D}}) \tag{6}
\end{equation*}
$$

are both weight-preserving bijections.
Since $d(v)(v \neq u)$ is the same for $\mathscr{D}$ and $\widetilde{\mathscr{D}}$ and $d(u)=$ $d\left(u_{1}\right)=d\left(u_{2}\right)=\cdots=d\left(u_{r}\right)$, we know that $W_{k}(\mathscr{D})=W_{k}(\widetilde{\mathscr{D}})$ for $k \geq 1$, and the invariance of $W_{\mathscr{D}}(t)$ follows.

Finally, we notice that $W_{k}(\mathscr{D})=\operatorname{tr}\left(O M^{k}\right)$. Thus

$$
\begin{align*}
W_{\mathscr{D}}(t) & =\sum_{k \geq 1} \operatorname{tr}\left\{O \cdot(t M)^{k}\right\} \\
& =\operatorname{tr}\left\{O \cdot \sum_{k \geq 1}(t M)^{k}\right\}  \tag{7}\\
& =\frac{\operatorname{tr}(O \cdot \operatorname{adj}(I-t M))}{\operatorname{det}(I-t M)}-\operatorname{tr}(O) .
\end{align*}
$$

Example 5. Let $\mathscr{D}=(D, w)$ be a weighted digraph as in the left of Figure 1. $\widehat{D}$ is the opposite of $D$ (see the right of Figure 1), that is, the digraph obtained from $D$ by reversing the direction of all its edges. It is easy to see that $D$ and $\widehat{D}$ have the same outdegree sequence $\{1,1,2,2,3\}$. The weight of any edge $(u, v) \in E(D)$ or $(u, v) \in E(\widehat{D})$ is defined to be $1 / d(u)$. Since $W_{3}(\mathscr{D})=5 / 4$ and $W_{3}(\widehat{D})=19 / 9$, we know that $\widehat{\mathscr{D}}$ cannot be obtained from $\mathscr{D}$ by a sequence of in-splittings or reverse operations.

Let $P$ be a nonnegative matrix. $P$ is called row stochastic if the summation of each row equals 1 and column stochastic if the summation of each column equals $1 P$ is called double stochastic if it is row and column stochastic.

Definition 6. Let $P$ be a row-stochastic matrix and $t$ a real positive number. Let $\mathscr{D}$ be the weighted digraph with weight matrix $M=t P$. We define $K(\mathscr{D}, t)$ as

$$
\begin{equation*}
K(\mathscr{D}, t)=(1+t) \frac{\sum_{v} d(v) \sum_{F_{v}} w\left(F_{v}\right)}{\sum_{F} w(F)}, \tag{8}
\end{equation*}
$$

where $v$ runs over all vertices of $V(D), F$ runs over all spanning directed forests of $D$, and $F_{v}$ runs over all spanning directed forests including $v$ as a root.

In general, $K(\mathscr{D}, t)$ is not an invariant under state insplitting, but the following result shows that it indeed reflects some invariance.

Corollary 7. Let P be a row-stochastic matrix and t a real positive number. Let $\mathscr{D}$ be a weighted digraph with weight matrix $M=t P$. Then $K(\mathscr{D}, t)-K(\widetilde{\mathscr{D}}, t)$ is an integer independent of $t$.

Proof. Let $O=\left(o_{i, j}\right)$ be the outdegree matrix of $D$. Then we get by Lemma 1 that

$$
\begin{align*}
K(\mathscr{D}, t) & =(1+t) \frac{\sum_{v} d(v) \sum_{F_{v}} w\left(F_{v}\right)}{\sum_{F} w(F)} \\
& =(1+t) \frac{\operatorname{tr}(O \cdot \operatorname{adj}[I+(t I-M)])}{\operatorname{det}[I+(t I-M)]}  \tag{9}\\
& =\operatorname{tr}\left(O \cdot\left[I-\frac{t}{1+t} P\right]^{-1}\right)
\end{align*}
$$

Since $P$ is stochastic and $1 /(1+t) \in(0,1)$, we have

$$
\begin{equation*}
\left[I-\frac{t}{1+t} P\right]^{-1}=[I-r(t P)]^{-1}=\sum_{i \geq 0}(M)^{i} r^{i} \tag{10}
\end{equation*}
$$

where $r=1 /(1+t)$.
Therefore

$$
\begin{align*}
K(\mathscr{D}, t) & =\operatorname{tr}\left(O \cdot\left[I-\frac{t}{1+t} P\right]^{-1}\right) \\
& =\sum_{i \geq 0} \operatorname{tr}\left\{O(M)^{i}\right\} r^{i}  \tag{11}\\
& =W_{\mathscr{D}}(r)+\operatorname{tr}(O) .
\end{align*}
$$

By Theorem 4, we know that $W_{\mathscr{D}}(r)$ is an invariant under in-splitting; thus

$$
\begin{equation*}
K(\mathscr{D}, t)-K(\widetilde{\mathscr{D}}, t)=\operatorname{tr}(O)-\operatorname{tr}(\widetilde{O}) \in \mathbb{Z} \tag{12}
\end{equation*}
$$

The result follows.
Lind and Tuncel defined a spanning tree invariant $\tau(\mathscr{D})$ for Markov shifts in [3] as follows:

$$
\begin{equation*}
\tau(\mathscr{D})=\sum_{T} w(T) \tag{13}
\end{equation*}
$$

Here the weight matrix $P$ of $\mathscr{D}$ is an irreducible row-stochastic matrix, and $T$ runs over all spanning trees of $\mathscr{D}$.

By considering the outdegree matrix as in Definitions 3 and 6 , we can define a new spanning tree invariant as

$$
\begin{equation*}
\tau_{d}(\mathscr{D})=\sum_{T} d(T) w(T) \tag{14}
\end{equation*}
$$

where $T$ is as above, and $d(T)$ denotes the outdegree of the root of $T$.

Corollary 8. $\tau_{d}(\mathscr{D})$ is an invariant under in-splitting.

Proof. Let $P$ be the weight matrix of $\mathscr{D}$ and thus row stochastic as in [3]. By the Matrix-Tree theorem (Theorem 2 in [8]), we have

$$
\begin{align*}
\tau_{d}(\mathscr{D}) & =\operatorname{tr}(O \cdot \operatorname{adj}[I-P]) \\
& =\lim _{t \rightarrow 1^{-}}\left\{\operatorname{det}[I-t P] \cdot \operatorname{tr}\left(O \cdot[I-t P]^{-1}\right)\right\} \\
& =\lim _{t \rightarrow 1^{-}}\left\{\operatorname{det}[I-t P] \cdot \sum_{i \geq 0} \operatorname{tr}\left(O P^{i}\right) t^{i}\right\}  \tag{15}\\
& =\lim _{t \rightarrow 1^{-}}\left\{\operatorname{det}[I-t P] \cdot\left(W_{\mathscr{D}}(t)+\operatorname{tr}(O)\right)\right\}
\end{align*}
$$

By Theorem 4, we know that $W_{\mathscr{D}}(t)$ is an invariant under state in-splitting. it is also well known that $\operatorname{det}[I-t P]$ is an invariant under state splitting. Therefore

$$
\begin{equation*}
\tau_{d}(\mathscr{D})-\tau_{d}(\widetilde{\mathscr{D}})=\lim _{t \rightarrow 1^{-}}\{\operatorname{det}[I-t P] \cdot(\operatorname{tr}(O)-\operatorname{tr}(\widetilde{O}))\} \tag{16}
\end{equation*}
$$

Since $\operatorname{tr}(O)-\operatorname{tr}(\widetilde{O})$ is a constant and $\lim _{t \rightarrow 1^{-}} \operatorname{det}[I-t P]=0$, we have

$$
\begin{equation*}
\tau_{d}(\mathscr{D})-\tau_{d}(\widetilde{\mathscr{D}})=0 \tag{17}
\end{equation*}
$$

The result follows.
Let $\mathscr{D}=(D, w)$ be a weighted digraph. The outweighted line digraph $L^{+}(\mathscr{D})=\left(L(D), w^{+}\right)$of $\mathscr{D}$ is a weighted digraph defined in the following way: the vertex set of $L(D)$ is $E(D) ;((u, v),(x, y)) \in E(L(D))$ if and only if $v=x$, and in this case, $w^{+}(((u, v),(x, y)))=w(x, y)$. Similarly, if we let $w^{-}(((u, v),(x, y)))=w(u, v)$ in the above definition, then we get the in-weighted line digraph $L^{-}(\mathscr{D})=$ $\left(L(D), w^{-}\right)$. Galeana-Sánchez and Gómez show that $L^{+}(\mathscr{D})$ can be obtained by sequences of state in-splittings from $\mathscr{D}$ (see Proposition 2.2 in [9], which has a small typo there by stating $L^{-}(\mathscr{D})$ can be obtained by sequences of state insplittings). Now the following conclusion is an immediate result of Corollary 8.

Corollary 9. $\tau_{d}(\mathscr{D})$ is an invariant under out-weighted line digraph operation.

## 3. The State Out-Splitting Case

Let $P$ be a row-stochastic matrix. Let $\mathscr{D}=(D, W)$ be the weighted digraph with weight matrix $W=P$. We first give the definition of state out-splitting, which is a little more complicated than the case of state in-splitting. Readers can refer to [3] for more details.

Definition 10. For a vertex $u$ of $D$, let $E^{* u}$ denote the set of edges of $D$ with initial end $u$. The state out-splitting of $\mathscr{D}$ at $u$ induces a new weighted digraph $\widetilde{\mathscr{D}}^{*}=\left(\widetilde{D}^{*}, \widetilde{w}\right)$ in the following way: let $\mathcal{S}^{*}=\left\{S_{1}^{*}, S_{2}^{*}, \ldots, S_{r}^{*}\right\}$ be a partition of $E^{* u}$. Let $q_{i}$ denote the sum of the weights of edges in $S_{i}^{*}$. The vertex set of the new digraph is $V\left(\widetilde{D}^{*}\right)=(V(D) \backslash$ $\{u\}) \bigcup\left\{u_{1}, u_{2} \ldots, u_{r}\right\}$. The edge set and weight of $\widetilde{\mathscr{D}}^{*}$ are defined as follows.
(i) For $x, y \in V(D) \backslash\{u\},(x, y) \in E\left(\widetilde{D}^{*}\right)$ if and only if $(x, y) \in E(D)$ and in this case $\widetilde{w}(x, y)=w(x, y)$.
(ii) For $y \in V(D) \backslash\{u\},\left(u_{i}, y\right) \in E\left(\widetilde{D}^{*}\right)$ if and only if $(u, y) \in S_{i}^{*}$ and in this case $\widetilde{w}\left(u_{i}, y\right)=w(u, y) / q_{i}$.
(iii) For $x \in V(D) \backslash\{u\},\left(x, u_{i}\right) \in E\left(\widetilde{D}^{*}\right)$ if and only if $(x, u) \in E(D)$ and in this case $\widetilde{w}\left(u_{i}, x\right)=q_{i} w(x, u)$.
(iv) If $(u, u) \in S_{i}^{*}$, then $\left(u_{i}, u_{j}\right) \in E\left(\widetilde{D}^{*}\right)$, for $j=1,2, \ldots, r$, and in this case $\widetilde{w}\left(u_{i}, u_{j}\right)=w(u, u) q_{j} / q_{i}$.

In the definition of $W_{k}(\mathscr{D})$ and $W_{D}(t)$, by replacing outdegrees with indegrees, we get $W_{k}^{*}(\mathscr{D})$ and $W_{\mathscr{D}}^{*}(t)$; that is,

$$
\begin{gather*}
W_{k}^{*}(\mathscr{D})=\sum_{v} d^{*}(v) \sum_{C \in C_{v}^{k}} w(C), \\
W_{\mathscr{D}}^{*}(t)=\sum_{k \geq 1} W_{k}^{*}(\mathscr{D}) t^{k}, \tag{18}
\end{gather*}
$$

where $d^{*}(v)$ is the indegree of $v$.
Theorem 11. Let $P$ be a row-stochastic matrix. Let $\mathscr{D}$ be the weighted digraph with weight matrix $P$. Then $W_{\mathscr{D}}^{*}(t)$ is an invariant under state out-splitting, and can be computed as

$$
\begin{equation*}
W_{\mathscr{D}}^{*}(t)=\frac{\operatorname{tr}\left(O^{*} \cdot \operatorname{adj}(I-t P)\right)}{\operatorname{det}(I-t P)}-\operatorname{tr}\left(O^{*}\right), \tag{19}
\end{equation*}
$$

where $O^{*}$ is the indegree matrix of $D$.
Proof. We just need to prove the invariance of $W_{k}^{*}(\mathscr{D})$ for $k \geq 1$. Without loss of generality, if there is a loop at vertex $u$, we assume that it belongs to $S_{1}^{*}$, where $\mathcal{S}^{*}=\left\{S_{1}^{*}, S_{2}^{*}, \ldots, S_{r}^{*}\right\}$ denotes the partition of $E^{* u}$ as in the definition of state outsplitting.

We define the mapping

$$
\begin{equation*}
\varphi: \bigcup_{v \in V(D)} C_{v}^{k}(\mathscr{D}) \longrightarrow \bigcup_{v \in V(\widetilde{D})} C_{v}^{k}\left(\widetilde{\mathscr{D}}^{*}\right) \tag{20}
\end{equation*}
$$

in the following way: for a closed walk $C$ of $\mathscr{D}$ with length $k$, if $C=(u, u, \ldots, u)$, then $\varphi(C)=\left(u_{1}, u_{1}, \ldots, u_{1}\right)$; otherwise, we replace each maximum path of $C$ of the form ( $u$, $u, \ldots, u, u, v)(v \neq u)$ with $\left(u_{1}, u_{1}, \ldots, u_{1}, u_{i}, v\right)$ if $(u, v) \in S_{i}^{*}$. By the definition of state out-splitting, it is not difficult to prove that

$$
\begin{equation*}
\varphi: C_{v}^{k}(\mathscr{D}) \longrightarrow C_{v}^{k}\left(\widetilde{D}^{*}\right) \tag{21}
\end{equation*}
$$

where $v \neq u$, and

$$
\begin{equation*}
\varphi: C_{u}^{k}(\mathscr{D}) \longrightarrow \bigcup_{i=1}^{r} C_{u_{i}}^{k}\left(\widetilde{\mathscr{D}}^{*}\right) \tag{22}
\end{equation*}
$$

are both bijections.
We now prove that they are also weight-preserving. In fact, if $C=(u, u, \ldots, u)$, then $w(C)=w(\varphi(C))$, since $\widetilde{w}\left(u_{i}\right.$, $\left.u_{i}\right)=w(u, u)$. On the other hand, for any walk of $C$ of the form $S=(r, u, \ldots, u, u, v)(v \neq u)$, we have $w(S)=w(r$, u) $w(u, u)^{k} w(u, v)$.
(1) If $(u, v) \in S_{1}^{*}$, we have

$$
\begin{align*}
w(\varphi(S)) & =w\left(r, u_{1}\right) w\left(u_{1}, u_{1}\right)^{k} w\left(u_{1}, v\right) \\
& =\frac{q_{1} w(r, u) w(u, u)^{k} w(u, v)}{q_{1}}  \tag{23}\\
& =w(S) .
\end{align*}
$$

(2) If $(u, v) \in S_{i}^{*}(i \neq 1)$ and $k \geq 1$, we have

$$
\begin{align*}
w(\varphi(S)) & =w\left(r, u_{1}\right) w\left(u_{1}, u_{1}\right)^{k-1} w\left(u_{1}, u_{i}\right) w\left(u_{i}, v\right) \\
& =\frac{q_{1} w(r, u) w(u, u)^{k-1}\left(w(u, u) q_{i} / q_{1}\right) w(u, v)}{q_{i}} \\
& =w(S) . \tag{24}
\end{align*}
$$

(3) If $(u, v) \in S_{i}^{*}(i \neq 1)$ and $k=0$, we have

$$
\begin{align*}
w(\varphi(S)) & =w\left(r, u_{i}\right) w\left(u_{i}, v\right) \\
& =\frac{q_{i} w(r, u) w(u, v)}{q_{i}}  \tag{25}\\
& =w(S)
\end{align*}
$$

Thus the maps above are weight preserving. Since $d^{*}(v)(v \neq u)$ is the same for $\mathscr{D}$ and $\widetilde{\mathscr{D}}^{*}$, and $d^{*}(u)=d^{*}\left(u_{1}\right)=$ $d^{*}\left(u_{2}\right)=\cdots=d^{*}\left(u_{r}\right)$, we know that $W_{k}^{*}(\mathscr{D})=W_{k}^{*}\left(\widetilde{D}^{*}\right)$, for $k \geq 1$, and the invariance of $W_{\mathscr{D}}^{*}(t)$ follows.

The proof of the equality is similar to that of Theorem 4.

Similarly, we can define $\tau_{d}^{*}(\mathscr{D})$ and prove that it is also an invariant under state out-splitting on the basis of the above result.

Now, we consider some weighted digraphs from [10] in the following two examples.

Example 12. The weight matrices of two weighted digraphs are as follows:

$$
A=\left[\begin{array}{ccc}
\frac{3}{8} & \frac{1}{2} & \frac{1}{8}  \tag{26}\\
0 & \frac{4}{5} & \frac{1}{5} \\
\frac{2}{7} & \frac{4}{7} & \frac{1}{7}
\end{array}\right], \quad B=\left[\begin{array}{ccc}
\frac{1}{7} & 0 & \frac{6}{7} \\
\frac{5}{56} & \frac{3}{8} & \frac{15}{28} \\
\frac{2}{15} & \frac{1}{15} & \frac{4}{5}
\end{array}\right] .
$$

By some computation, we get that $W_{\mathscr{A}}(1 / 2)=1316 / 471$, $W_{\mathscr{B}}(1 / 2)=1615 / 471$, and $W_{\mathscr{A}}^{*}(1 / 2)=W_{\mathscr{B}}^{*}(1 / 2)=1559 / 471$. Thus $B$ cannot be archived by a sequence of in-splittings or reverse operations, but may be archived by a sequence of outsplittings or reverse operations.

Example 13. The weight matrices of three weighted digraphs are as follows:

$$
A=\left[\begin{array}{ll}
\frac{1}{3} & \frac{2}{3}  \tag{27}\\
\frac{1}{3} & \frac{2}{3}
\end{array}\right], \quad B=\left[\begin{array}{cc}
\frac{1}{3} & \frac{2}{3} \\
\frac{2}{3} & \frac{1}{3}
\end{array}\right], \quad C=\left[\begin{array}{cc}
\frac{2}{3} & \frac{1}{3} \\
\frac{1}{3} & \frac{2}{3}
\end{array}\right] .
$$

By some computation, we get that $\tau_{d}^{*}(A)=2=\tau_{d}(A)$, $\tau_{d}^{*}(B)=8 / 3=\tau_{d}(B), \tau_{d}^{*}(C)=4 / 3=\tau_{d}(C)$. Thus for any pair of them, we cannot get one from the other and by a sequence of in-splittings or reverse operations either nor by a sequence of out-splittings or reverse operations.

## 4. Invariants for Weighted Digraphs with Double-Stochastic Matrices

Let $\mathscr{D}=(D, P)$ be a weighted digraph. If the weight matrix $P$ is column stochastic, the weight distribution after state out-splitting can be defined in an easier way, that is, without multiplying by the coefficients about $q_{i}$ in Definition 10. Under this definition, we can get that $\tau_{d}^{*}(\mathscr{D})$ is still an invariant under state out-splitting, the proof of which is similar to that of Corollary 8 . We also know from [9] that the in-weighted line digraph can be obtained by a sequence of such state out-splittings, so the following result is immediate.

Corollary 14. Let $\mathscr{D}=(D, P)$ be a weighted digraph. If the weight matrix $P$ is column stochastic, then $\tau_{d}^{*}(D)$ is an invariant under in-weighted line digraph operation.

Especially, if the weight matrix is doubly stochastic, we have the following result.

Corollary 15. Let $\mathscr{D}=(D, P)$ be a weighted digraph. If the weight matrix $P$ is doubly stochastic, then $\tau_{d}\left(L^{+}(\mathscr{D})\right)=$ $\tau_{d}^{*}\left(L^{-}(\mathscr{D})\right)$.

Proof. Since $P$ is doubly stochastic, we have by Corollary 8 that

$$
\begin{equation*}
\tau_{d}\left(L^{+}(\mathscr{D})\right)=\tau_{d}(\mathscr{D})=\operatorname{tr}(O \cdot \operatorname{adj}[I-P]) \tag{28}
\end{equation*}
$$

and by Corollary 14 that

$$
\begin{equation*}
\tau_{d}^{*}\left(L^{-}(\mathscr{D})\right)=\tau_{d}^{*}(\mathscr{D})=\operatorname{tr}\left(O^{*} \cdot \operatorname{adj}[I-P]\right) \tag{29}
\end{equation*}
$$

By Matrix-Tree theorem (Theorem 2 in [8]), we know that both $\operatorname{adj}[I-P]$ and $\operatorname{adj}\left[I-P^{\prime}\right]=(\operatorname{adj}[I-P])^{\prime}$ are rowconstant matrices, where $P^{\prime}$ is $P$ transposed. Thus adj $[I-P]$ is a constant matrix. Since the sum of indegrees is equal to that of outdegrees, the result follows.

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## References

[1] B. P. Kitchens, Symbolic Dynamics: One-Sided, Two-Sided and Countable State Markov Shifts, Springer, Berlin, Germany, 1998.
[2] D. Lind and B. Marcus, An Introduction to Symbolic Dynamics and Coding, Cambridge University Press, Cambridge, UK, 1995.
[3] D. Lind and S. Tuncel, "A spanning tree invariant for Markov shifts," in Codes, Systems, and Graphical Models, vol. 123 of The IMA Volumes in Mathematics and Its Applications, pp. 487-497, Springer, New York, NY, USA, 2001.
[4] R. B. Bapat, "On a conjecture concerning spanning tree invariants and loop systems," Linear Algebra and Its Applications, vol. 433, no. 8-10, pp. 1642-1645, 2010.
[5] R. Gómez, "On mean recurrence times of Markov chains and spanning tree invariants," Linear Algebra and Its Applications, vol. 433, no. 11-12, pp. 1714-1718, 2010.
[6] R. Gómez and J. M. Salazar, "Spanning tree invariants, loop systems and doubly stochastic matrices," Linear Algebra and Its Applications, vol. 432, no. 2-3, pp. 556-565, 2010.
[7] B. Bollobás, Modern Graph Theory, vol. 184, Springer, New York, NY, USA, 1998.
[8] P. Yu. Chebotarev and E. V. Shamis, "The Matrix-Forest theorem and measuring relations in small social groups," Automation and Remote Control, vol. 58, no. 9, pp. 1505-1514, 1997.
[9] H. Galeana-Sánchez and R. Gómez, "Monochromatic paths on edge colored digraphs and state splittings," Advances and Applications in Discrete Mathematics, vol. 4, no. 1, pp. 33-51, 2009.
[10] R. Cowen and E. M. Lungu, "When are two Markov chains the same?" Quaestiones Mathematicae, vol. 23, no. 4, pp. 507-513, 2000.

