Transport and Vulnerability in River Deltas: A Graph-Theoretic Approach

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Maintaining a sustainable socio-ecological state of a river delta requires delivery of material and energy fluxes to its body and coastal zone in a way that avoids malnourishment that would compromise system integrity. We present a quantitative framework for studying delta topology and transport based on representation of a deltaic system by a rooted directed acyclic graph. Applying results from spectral graph theory allows systematic identification of the upstream and downstream subnetworks for a given vertex, computing steady flux propagation in the network, and finding partition of the flow at any channel among the downstream channels. We use this framework to construct vulnerability maps that quantify the relative change of sediment and water delivery to the shoreline outlets in response to possible perturbations in hundreds of upstream links. This enables us to evaluate which links (hotspots) and what management scenarios would most influence flux delivery to the outlets. The results can be used to examine local or spatially distributed delta interventions and develop a system approach to delta management.

Deltas are landforms with channels that deliver water, sediment and nutrient fluxes from rivers to oceans or inland water bodies via multiple pathways. These systems evolve naturally by maintaining the balance between subsidence due to compaction and new land formation due to sediment deposition from the river upstream [1]. This dynamic interaction results in a low-relief terrain with slopes as low as $1.0 \times 10^{-5}$ [2].

Deltas are highly productive regions supporting extensive agriculture, diverse ecosystems, and containing natural resources such as hydrocarbon deposits. More than half a billion people reside in deltas with over 300 million living in the Ganges-Brahmaputra-Meghna, Yangtze and Nile alone [3, 4]. Unfortunately, many deltas are vulnerable to both natural and anthropogenic drivers, and are predicted to be in danger of collapsing within the 21st century [5]. Any alteration in the delta network can provoke physical (channel morphology), biological (ecosystems) and socio-economic changes. For instance, rapid sea-level rise exacerbates land loss inducing saltwater intrusion upstream and loss of ecosystem habitat. On the other hand, demands on water and energy from upstream are satisfied by dams and divergence structures, while multiple dykes, embankments and sluice gates are constructed downstream mainly for irrigation purposes and to control floods and salinity intrusion. These human-engineered structures result in unintended consequences that prevent the growth of deltas and disrupt their natural ecosystem dynamics. The rising sea-level, in concert with reduction in aggradation and accelerated compaction, is putting many deltas in peril [3, 4].

Recent works have focused mainly on modeling delta growth and evolution [6-11] and developing quantitative metrics to describe delta morphology [12, 13]. At the same time, formal methodologies for studying deltas’ topology and dynamic processes operating on them are still lacking. Our study contributes to this direction. Specifically, we conceptualize a delta as a rooted acyclic directed graph and use its weighted adjacency matrix to (i) identify the steady flow along the delta channels, (ii) detect upstream (contributing) and downstream (nourishing) subnetworks for a given channel, (iii) find the partition of flow from a given channel to any collection of downstream nodes at steady state, and (iv) examine how perturbations at upstream parts of the system propagate downstream. The developed framework is illustrated by examining the vulnerability of the downstream (shoreline) outlets to flux reduction in upstream channels and building vulnerability maps that can facilitate assessment of delta development scenarios.

Recall that a graph $\mathcal{G} = (V,E)$ is a collection of vertices $V = \{v_i\}, i = 1, \ldots, N$ and edges $E = \{(uv)\}$, $u,v \in V$, where the notation $(uv)$ signifies that the edge connects the vertices $u$ and $v$. A graph is called directed, or digraph, if its edges have directions, that is, edge pairs $(uv)$ are ordered. A digraph is called acyclic if there are no directed paths from a vertex to itself. A digraph is called rooted if there is a vertex $r$ such that there exists a directed path from $r$ to any other vertex in the graph. A digraph can be uniquely specified by its (asymmetric) adjacency matrix $A$ such that $A(v,u) = 1$ if there is an edge $(uv)$ and $A(v,u) = 0$ otherwise. Sometimes, edges are given weights $w_{uv}$. In this case the graph is specified by its weighted adjacency matrix $W$ with non-zero elements...
nontrivial steady state. To avoid this problem we consider a cycled version of the network, where the outlets (leaves) directly drain their entire flux to the apex (root). The weighted adjacency matrix for this new network is denoted by \( \tilde{W} \). Hence, we seek a solution \( F = (F_1, \ldots, F_N)^T \) (where \( x^T \) is the transpose of \( x \)) of the system

\[
F_i = \sum_j \tilde{w}_{ji} F_j, \quad i = 1, \ldots, N. \tag{1}
\]

This can be written in matrix notation as \( F = \tilde{W} F \), or \( (I_N - \tilde{W})F = 0_N \), where \( I_N \) is the \( N \times N \) identity matrix, and \( 0_N \) is an \( N \times 1 \) vector of zeros. In other words, we need to identify the null space of the matrix \( L_{SF} = I - \tilde{W} \). We also notice that \( I_N \) is the out-degree matrix for \( \tilde{W} \), so \( L_{SF} \) is a graph Laplacian for \( \tilde{W} \). The steady flow at an edge \( (uv) \) is given by \( F_{(uv)} = F_u w_{uv} \). It is clear from the flux interpretation that there exists a unique solution to this problem.

**Proposition 1** (Steady flow). The steady flow \( F \) through a rooted acyclic digraph \( G \) with weighted cycled adjacency matrix \( \tilde{W} \) is given (up to a scalar factor) by the eigenvector that spans the null space of the graph Laplacian:

\[
\text{null} \left( L_{SF} \right) = \text{null} \left( I - \tilde{W} \right) = \left\{ x : (I - \tilde{W})x = 0_N \right\}.
\]

Next, we use the results of \( [14, 15] \) to identify the subnetwork that participates in draining fluxes from the apex to a given vertex \( u \) (contributing network) and the subnetwork that drains fluxes from \( u \) to the outlets (nourishing network) \( [12] \). Recall that a reachable set \( R(i) \) for vertex \( i \) in a digraph \( G \) is the collection of all vertices \( j \) such that there exists a directed path from \( i \) to \( j \). The set \( R \) is called a reach if it is a maximal reachable set, that is \( R = R(i) \) for some \( i \) and there is no such \( j \) that \( R(i) \subset R(j) \). The exclusive part of a reach \( R_i \) is defined as \( H_i = R_i \setminus \cup_{j \neq i} R_j \). The common part of \( R_i \) is defined as \( C_i = R_i \setminus H_i \). Caughman and Veerman \( [15] \) prove the following theorem.

**Theorem 1** (Reaches of a digraph). Let \( A \) be the \( N \times N \) adjacency matrix for \( G \) and \( D \) be the \( n \)-degree matrix for \( A \), that is the diagonal \( N \times N \) matrix with diagonal elements taken from \( A1_N \). Then the nullspace of the Laplacian \( L = D - A \) has a basis \( \gamma_i \) in \( \mathbb{R}^N \) whose elements satisfy: (i) \( \gamma_i(v) = 0 \) for \( v \notin R_i \); (ii) \( \gamma_i(v) = 1 \) for \( v \in H_i \); (iii) \( \gamma_i(v) \in (0, 1) \) for \( v \in C_i \); \( \gamma_i \) forms a basis for \( \text{null} (L) \).

Suppose that a delta system is represented by a rooted acyclic digraph \( G \). Consider now the same stream topology with reversed flux directions; the new network is specified by the acyclic directed graph \( G^R \) with adjacency matrix \( A^T \). Each outlet \( i = 1, \ldots, k \) of the initial delta generates a reach \( R_i \) within \( G^R \). The exclusive part of \( R_i \) consists of the vertices that in \( G \) drain exclusively to the
outlet $i$. The common part of $R_i$ consists of the vertices that in $G$ also drain to at least one other outlet. It is easily seen that there are no other reaches in $G^R$. Each outlet $i$ of the initial delta belongs to the exclusive part of the respective reach in $G^R$. The apex of the initial delta belongs to all the reaches of $G^R$. Let $B$ denote the in-degree matrix for $A^T$, that is the diagonal $N \times N$ matrix with diagonal elements taken from $A^T 1_N$. Consider the Laplacian $Q = B - A^T$ of the digraph $G^R$. A basis for the nullspace of $Q$ is described by Theorem 1. The subnetwork of $G$ that drains from the apex to the outlet $i$ corresponds to the unique eigenvector of the $Q$-nullspace with $\gamma(i) = 1$. The nonzero elements of this eigenvector identify the vertices that participate in the subnetwork.

Notably, if we consider a flow along the directed edges of $G$ with equal distribution of the parental flux among the offsprings, then the elements of eigenvectors from the $Q$-nullspace allow the following interpretation. The element $\gamma_i(v)$ equals the proportion of the flux at vertex $v$ that drains to outlet $i$ in the original network $G$. This statement is trivial for the outlets of $G$, which always belong to the exclusive part of respective reaches. To prove the statement for the rest of vertices, observe that $B^T \gamma_i = A^T \gamma_i$, which means that the element $v$ of the eigenvector $\gamma_i$ equals the average of its parental elements in $G^R$. The results of [13] apply as well to general stochastic adjacency matrices, which allows one to generalize the above discussion to the situation with unequal distribution of parental flux among the offsprings.

**Corollary 1** (Contributing network). Consider flow along the edges of a rooted acyclic graph $G$ specified by a weighted adjacency matrix $W$ whose elements $w_{uv}$ represent the proportion of flux at parental vertex $u$ that drains to offspring vertex $v$. Let $Q = B - W^T$ be the weighted graph Laplacian, with $B$ being the in-degree matrix for $W^T$. Then the nullspace of $Q$ has a basis $\gamma_i$ in $\mathbb{R}^N$ such that $\gamma_i(v)$ equals the proportion of flux at vertex $v$ that drains to outlet $i$. In particular, $\gamma_i(v) \neq 0$ if and only if outlet $i$ receives fluxes from vertex $v$.

Corollary 1 is readily applied to finding the subnetwork that participates in draining fluxes to any chosen vertex, not necessarily an outlet. For that, one needs to make the examined vertex an outlet, by disconnecting it from its offspring. The modified weighted adjacency matrix is then used to obtain the result.

To identify the nourishing network for a node $u$, we first make $u$ an apex by disconnecting it from the parents. The identification is done using the following result.

**Corollary 2** (Nourishing network). Consider flow along the edges of an acyclic graph $G$ with $k$ roots, specified by a weighted adjacency matrix $W$ whose elements $w_{uv}$ represent the proportion of flux at parental vertex $u$ that drains to offspring vertex $v$. Let $L = D - W$ be the weighted graph Laplacian with $D$ being the in-degree matrix for $W$. Then the nullspace of $L$ has a basis $\gamma_i$, $i = 1, \ldots, k$ in $\mathbb{R}^N$ such that $\gamma_i(v) \neq 0$ if and only if vertex $v$ receives fluxes from root $i$.

![FIG. 2. Steady state flux (a,c) and number of outlets (b,d) that a given link contributes to. (a,b) Wax Lake delta: The distribution of flux among the immediate downstream links is proportional to the channel width. (c,d) Niger delta: The flux is distributed equally among the immediate downstream links. The flux at the apex is normalized to $F = 1$.](image)
result is comparable to the synthetic sediment flux distribution at the shoreline for the Wax Lake delta obtained by Edmonds et al. \[12\]. The steady state flux for the Niger delta is illustrated in Fig. 2(c). This delta has a single dominant outlet that receives 50% of the apex flux, with the second largest outlet receiving 12%. Figures 2(b),(d) show the number of outlets a given link contributes to in the two examined deltas. This plot highlights the relative importance of a link in the delta network: the hotspot (red) links affect many outlets while blue links only affect a single outlet. The hotspots can be interpreted as “highways of perturbation” since, even if the steady flux in the link is not high, the effect of the perturbation will be experienced by many shoreline outlets in the delta. Representative examples of the outlet contributing networks are highlighted in Fig. 3(a)-(d) (where the colors should be ignored for now). In the Wax Lake delta, most of the outlet contributing networks (18 out of 24, or 67%) have a single path connecting the delta apex to the shoreline outlet. On the contrary, in the Niger delta all the networks have multiple pathways.

Flux reduction is recurrent in deltas due to dams and impoundments. Flux reduction at edge \((uv)\) leads to flux reduction within the nourishing area of \(v\). The flux reduction is specified by adding a new outlet \(z\) to the vertex \(v\) and assigning new weights \(w_{uz} = w_{uv}^{\text{old}} (1 - \alpha)\) and \(w_{uv}^{\text{new}} = w_{uv}^{\text{old}} \alpha\), \(0 < \alpha < 1\). The reduced flux in the entire delta is computed using Proposition 1 with the updated weighted adjacency matrix. We are interested in identifying “vulnerable” links defined as the links whose flux reduction would cause the highest reduction at the outlets. For that we compute the flux reduction at the outlets caused by an \(\alpha = 0.4\) flux reduction at a given link, considered one-by-one. The results are illustrated in vulnerability maps for the representative subnetworks in Fig. 3. The links colored red, yellow, and blue represent high (\(r > 40\%\)), medium (\(20\% < r < 40\%\)) and low (\(r < 20\%\)) flux reduction at the shoreline outlet, respectively. In general, if a given link drains \(p \cdot 100\%, \; 0 \leq p \leq 1\), of its flux to a given outlet, and the steady flux \(f^{\text{old}}\) at the link is related to the steady flux \(g^{\text{old}}\) at the outlet as \(f^{\text{old}} = C g^{\text{old}},\; C > 0\), then \(\alpha\)-reduction at this link results in the outlet reduction \(g^{\text{new}} = f^{\text{old}} (1 - \alpha p C)\). Here, \(\alpha\) is a local characteristic of the link, while both \(C\) and \(p\) are spatially extended pairwise characteristics of the link and the examined outlet. We also notice that in multi-paths networks (Fig. 3 except (c)) the vulnerability might not be monotone along individual downstream channels. These observations support the necessity of a systematic, spatially-extended approach to studying the effects of link modifications.

In contrast to well-studied topology of tributary channel networks (networks that drain to a single outlet; e.g. \[18\]), the exploration of the topology of distributary channel networks (networks that originate from a single source and drain to multiple outlets) is still in its infancy. Yet, this topology defines the distribution of network fluxes and dictates how changes in a given part of a network propagate to the rest; it also paves the way to better understand the intricate self-organization of deltaic systems. We present here a framework for analyzing the topology of delta networks, and specifically for identifying the upstream and downstream subnetworks for any given vertex, computing steady-state flux propagation in the network, and performing vulnerability analysis by assessing parts of the network where a change would most significantly affect the downstream or shoreline fluxes. Notably, all the results follow directly from the spectral decomposition of the Laplacian for the graph representing the examined delta. Although we focus on the steady state topology of a delta that does not directly incorporate the dynamic evolution of the channel morphology or its topology, extension of the framework to incorporate time-evolving adjacency matrix is possible. The proposed framework can form a basis for delta network topology classification and for defining comparative vulnerability metrics among different deltas as well as for the same delta under natural and/or human-induced changes.

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