

# Orienting an Edge-Bi-Weighted Graph to Minimize the Heaviest Path

# 1 Introduction

A *edge bi-weighted graph* is an undirected graph  $G = (V, E)$ , such that each edge  $\{u, v\} \in E$  has a pair of (possibly different) weights  $w(u, v)$  and  $w(v, u)$  associated with its two possible orientations  $(u, v)$  and  $(v, u)$ , respectively. An *orientation*  $\vec{O}$  of  $G$  creates a directed graph  $\vec{G} = \vec{O}(G)$  from  $G$  by selecting for each undirected edge  $\{u, v\} \in E$  exactly one of its two possible orientations. Denote by  $\mathcal{O}(G)$  the set of all orientations of  $G$ .

The problem we address is the following.

**Input:** An edge bi-weighted graph  $G$ .

**Output:** An orientation of  $G$  that minimizes the weight of the heaviest of the resulting simple directed paths.

Actually, there remains an ambiguity in the specification of the cost function because it fails to specify whether the minimum is taken only over maximal simple paths or over all simple subpaths. To distinguish between these two possibilities we define two measures.

**Definition 1.1.** *Given a directed path  $\vec{P} = \langle v_0, \dots, v_n \rangle$  let*

$$h_m(\vec{P}) = \sum_{k=0}^{n-1} w(v_k, v_{k+1}), \quad h_s(\vec{P}) = \max_{0 \leq i < j \leq n-1} \sum_{k=i}^j w(v_k, v_{k+1}). \quad (1)$$

*We also extend these measures to an oriented graph  $\vec{G}$*

$$h_m(\vec{G}) = \max\{h_m(\vec{P}) : \vec{P} \text{ is a maximal simple path in } \vec{G}\}, \quad (2)$$

$$h_s(\vec{G}) = \max\{h_s(\vec{P}) : \vec{P} \text{ is a simple path in } \vec{G}\}. \quad (3)$$

*The corresponding cost functions for orienting an undirected graph are*

$$H_m(G) = \min\{h_m(\vec{O}(G)) : \vec{O} \in \mathcal{O}(G)\}, \quad (4)$$

$$H_s(G) = \min\{h_s(\vec{O}(G)) : \vec{O} \in \mathcal{O}(G)\}. \quad (5)$$

To illustrate the differences between the two cost functions consider  $\vec{P} = \langle v_0, v_1, v_2, v_3 \rangle$ , with  $w(v_0, v_1) = 2, w(v_1, v_2) = -3, w(v_2, v_3) = 6$ . Then  $h_m(\vec{P}) = 5, h_s(\vec{P}) = 6$ .

A very useful property of  $h_s$  is that it is monotone: the cost of a graph is never less than the cost of a subgraph.

**Lemma 1.1.** *Given any edge bi-weighted graph  $G$ ,  $H_s(G) \geq H_s(G')$  for any subgraph  $G'$  of  $G$ .*

**Proof:** It is clear from the definition (1) of  $h_s$  that  $h_s(\vec{P}) \geq h_s(\vec{P}')$  for any subpath  $\vec{P}'$  of  $\vec{P}$ .  $\square$

This property sets  $h_m$  apart from  $h_s$ , as the above example shows:  $h_m(\vec{P}) = 5 < h_m(\vec{P}') = 6$  for the subgraph  $\vec{P}' = \langle v_2, v_3 \rangle$  of  $\vec{P}$ .

However, if all weights are non-negative  $h_m$  too is monotonic. In fact, in that case the two definitions coincide.

**Lemma 1.2.** *Suppose that  $G$  is an edge bi-weighted graph all of whose weights are non-negative. Then  $h_s(\vec{G}) = h_m(\vec{G})$  for any orientation  $\vec{G}$  of  $G$ . In particular,  $H_s(G) = H_m(G)$ , and  $\vec{O}$  is an optimal orientation of  $G$  with respect to  $H_s$  if and only if it is an optimal orientation with respect to  $H_m$ .*

**Proof:** From the definitions of  $h_m$  and  $h_s$  it is clear that for any path graph  $P$   $h_m(P) \leq h_s(P)$ , and that the inequality is in fact an equality if all the weights are non-negative. This implies  $H_s(G) = H_m(G)$ .

Clearly any orientation of  $G$  that is optimal with respect to  $h_m$  is also optimal with respect to  $h_s$ .  $\square$

**Conjecture:** *Given an edge bi-weighted graph  $G$ , any orientation that is optimal with respect to  $h_m$  is also optimal with respect to  $h_s$ .*

In the following sections we will consider the two problems for various classes of graphs.

**Problem 1.1** (Minimizing the heaviest subpath - HS). *Given a class of edge bi-weighted graphs find an algorithm to compute, for any  $G$  in the class,*

$$H_s(G) = \min_{\vec{O} \in \mathcal{O}(G)} h_s(\vec{G})$$

**Problem 1.2** (Minimizing the heaviest maximal path - HM). *Given a class of edge bi-weighted graphs find an algorithm to compute, for any  $G$  in the class,*

$$H_m(G) = \min_{\vec{O} \in \mathcal{O}(G)} h_m(\vec{G})$$

## 2 Algorithms for paths

A *path* in a directed graph  $\vec{G} = (V, E)$  is a series of vertices  $P = \langle v_0, v_1, \dots, v_m \rangle$  such that for every  $0 \leq i < m$ ,  $(v_i, v_{i+1}) \in E$ . Such a path is *maximal* if for every  $u \in V$ ,  $(u, v_0), (v_m, u) \notin E$ . The weight of a path is defined as the sum of path edge weights. Denote by  $h(\vec{G})$  and  $\tilde{h}(\vec{G})$  maximum weights of a path and a maximal path in  $\vec{G}$ , respectively. [SZ: how should cycles be handled?]