
Algorithm 1: A generalized McAulay-Quatieri peak-matching algorithm.

Input: the cost matrix C

Output: L tuples of indices Γ , or fewer if Δ_{MQ} exceeded

$\Gamma \leftarrow \emptyset$;

for $l \leftarrow 0$ **to** $L - 1$ **do**

$\Gamma_l = \arg \min_{[0, \dots, M_0 - 1] \times \dots \times [0, \dots, M_{K-1} - 1] \setminus \Gamma} C$;

if $\exists i, j \in \Gamma_l : \mathcal{D}(\theta_i, \theta_j) > \Delta_{MQ}$ **then**

return Γ

end

$\Gamma \leftarrow \Gamma \cup C_{\Gamma_l}$;

end

return Γ

smaller costs to be chosen in successive iterations. This algorithm does not allow for that. In other terms, the algorithm does not find a set of pairs that represent a globally minimal sum of costs. Furthermore, the algorithm does not scale well: assuming equal numbers of parameter sets in all frames, the search space grows exponentially with K . Nevertheless, the method is simple to implement, computationally negligible when K is small, and works well with a variety of audio signals such as speech [1] and music [9].

3. L BEST PATHS THROUGH A LATTICE VIA LINEAR PROGRAMMING (LP)

In this section we show how to find L paths through a lattice of K frames such that the sets of nodes on each path are disjoint. The k th frame of the lattice contains N_k nodes for a total of $M = \sum_{k=0}^{K-1} N_k$ nodes.

Similar to the McAulay-Quatieri method we define the cost Δ_{LP} as the limiting cost under which the connection between two nodes will be considered in the LP method.

The solution vector \mathbf{x} to the linear program shall indicate the presence of a connection between a pair of nodes by having an entry equal to 1 and otherwise have entries equal to 0. To enumerate the set of possible connection-pairs we define

$$\rho = \{(i, j) : \mathcal{D}(\theta_i, \theta_j) \leq \Delta_{LP}, 0 \leq i < M, 0 \leq j < M, i \neq j\} \quad (1)$$

The cost vector of the objective function is then

$$\mathbf{c}_\rho = \{D(\theta_i, \theta_j) \forall (i, j) \in \rho\} \quad (2)$$

and the length of \mathbf{c}_ρ is $\#\rho = \#\mathbf{c}_\rho = P$, in other words, P pairs of nodes. For convenience we define a bijective mapping $\mathcal{B} : \rho \rightarrow [0, \dots, M - 1]$ giving the index in \mathbf{x} of the pair $p \in \rho$. For the implementation considered in this paper, $\mathcal{D}(\theta_i, \theta_j) = \infty$ for all i, j not in adjacent frames and so P will be no larger than $(K - 1)N^2$ (assuming the same number of nodes N in each frame).

The total cost of the paths in the solution is then calculated through the inner product $\mathbf{c}_\rho^T \mathbf{x}$. To obtain \mathbf{x}^* that represents L disjoint paths we must place constraints on the structure of the solution. Some of the constraints presented in the following are redundant but the redundancies are kept for clarity; later we will show which constraints can be removed without changing the optimal solution \mathbf{x}^* .

All nodes in \mathbf{x}^* will have at most one incoming connection or otherwise no connections, a constraint that can be enforced through the following linear inequality: define $\mathbf{A}^I \in \mathbb{R}^{R_I \times P}$ with $R_I =$

$\sum_{k=1}^{K-1} N_k$, the number of nodes in all the frames excluding the first. We sum all the connections into the node $r_1 + N_0$ represented by the respective entry in \mathbf{x} through an inner product with the r_1 th row in \mathbf{A}^I and require that this sum be between 0 and 1, i.e.,

$$\mathbf{A}_{r_1, \mathcal{B}(p)}^I = \begin{cases} 1 & \text{if } p_1 = r_1 + N_0, 0 \leq r_1 < R_I, p \in \rho \\ 0 & \text{otherwise} \end{cases} \quad (3)$$

and

$$\mathbf{0} \leq \mathbf{A}^I \mathbf{x} \leq \mathbf{1} \quad (4)$$

Similarly, to constrain the number of outgoing connections into each node, we define $R_O = \sum_{k=0}^{K-2} N_k$ and $\mathbf{A}^O \in \mathbb{R}^{R_O \times P}$ with

$$\mathbf{A}_{r_O, \mathcal{B}(p)}^O = \begin{cases} 1 & \text{if } p_0 = r_O, 0 \leq r_O < R_O, p \in \rho \\ 0 & \text{otherwise} \end{cases} \quad (5)$$

and

$$\mathbf{0} \leq \mathbf{A}^O \mathbf{x} \leq \mathbf{1} \quad (6)$$

To forbid breaks in the paths it is required that the number of incoming connections into a given node equal the number of outgoing connections for the $R_B = \sum_{k=1}^{K-2} N_k$ nodes potentially having both incoming and outgoing connections.

$$\mathbf{A}_{r_B}^B = \mathbf{A}_{r_B}^B - \mathbf{A}_{r_B + N_0}^B \text{ for rows } 0 \leq r_B < R_B \quad (7)$$

and

$$\mathbf{A}^B \mathbf{x} = \mathbf{0} \quad (8)$$

Finally we ensure that there are L paths by counting the number of connections in each frame and constraining this sum to be L . We choose arbitrarily to count the number of outgoing connections by summing rows of \mathbf{A}^O into rows of $\mathbf{A}^C \in \mathbb{R}^{(K-1) \times P}$

$$\mathbf{A}_{r_C}^C = \sum_{k=a}^b \mathbf{A}_k^O \quad (9)$$

with $a = \sum_{j=0}^{r_C} N_j$ and $b = \sum_{j=0}^{r_C+1} N_j$ and

$$\mathbf{A}^C \mathbf{x} = L \mathbf{1} \quad (10)$$

As stated above, some of these constraints are redundant and can be removed. Indeed, we have $\mathbf{0} \leq \mathbf{x} \leq \mathbf{1}$, therefore we will always have $\mathbf{A}^I \mathbf{x} \geq 0$ and $\mathbf{A}^O \mathbf{x} \geq 0$. Furthermore, all but the last row of (10) can be seen as constructed from linear combinations of rows of (8) and the last row of (10) so we only require $\mathbf{A}_{K-2}^C \mathbf{x} = L$. Finally we always have $\mathbf{x} \leq \mathbf{1}$ because of the constraint that there be a maximum of 1 incoming and outgoing connection from each node.

The complete LP to find the L best disjoint paths through a lattice described by node connections ρ is then

$$\min_{\mathbf{x}} \mathbf{c}_\rho^T \mathbf{x}$$

subject to

$$\begin{aligned} \mathbf{G}\mathbf{x} &= \begin{bmatrix} \mathbf{A}^I \\ \mathbf{A}^O \\ -\mathbf{I} \end{bmatrix} \mathbf{x} \leq \begin{bmatrix} \mathbf{1} \\ \mathbf{1} \\ \mathbf{0} \end{bmatrix} \\ \mathbf{A}\mathbf{x} &= \begin{bmatrix} \mathbf{A}^B \\ \mathbf{A}_{K-2}^C \end{bmatrix} \mathbf{x} = \begin{bmatrix} \mathbf{0} \\ L \end{bmatrix} \end{aligned} \quad (11)$$

where \mathbf{I} is the identity matrix. A proof that the solution \mathbf{x}^* will have entries equal to either 0 or 1 can be found in [10, p. 167].