

# Path Color Switching

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# Problem Description

We want to generate sequences of musical “chords” with some known constraints as well as control on the complexity of the sequence.

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Spotify

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**Input** An oriented graph whose arcs are colored with a set of colors, two nodes of the graphs  $s$  and  $t$  and a length  $k$ .

**Output** Set single colors to edges to find a path of length  $k$  from  $s$  to  $t$  minimizing the number of color switch.

# Definitions & notations

*Color switch (CS)*: given two adjacent arcs  $a_1$  and  $a_2$  colored respectively with  $c_1$  and  $c_2$ , we have a color CS if  $c_1 \neq c_2$ .

# Definitions & notations

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$\mathcal{G} = (V, A)$ : A directed graph where  $V$  is the set of its nodes and  $A$  is the set of its arcs.

$\mathcal{C}$ : A finite set of colors.

$\mathcal{F}$ : The coloring function defined as  $\mathcal{F} : A \rightarrow 2^{\mathcal{C}}$ .

$\mathcal{P} = (v_1, \dots, v_k)$ : A path going from  $v_1$  to  $v_k$ .

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$w(\mathcal{P})$ : The cost of the path  $\mathcal{P}$  which is given by the sum of its CS.

# Problem decomposition

The problem can be decomposed into small parts:

- Minimize CS on paths;
- Minimize CS on graphs.

# Minimize CS on Paths

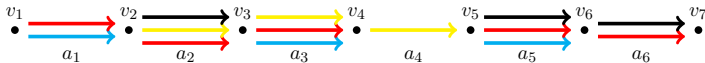


Figure: A path  $\mathcal{P}$

What is the color assignment minimizing  $w(\mathcal{P})$ ?



# Algorithm

Let  $\mathcal{P} = (a_1, \dots, a_k)$  a path

Let  $\mathcal{T} : A \rightarrow 2^{\mathcal{C}}$  a function such that:

- $\mathcal{T}(a_1) = \mathcal{F}(a_1)$
  - $\mathcal{T}(a_i) = \mathcal{F}(a_i) \cap \mathcal{T}(a_{i-1})$  if not empty else  $\mathcal{F}(a_i)$
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$\mathcal{H} : A \rightarrow \mathcal{C}$  the function minimizing  $w(\mathcal{P})$  such that:

- $\mathcal{H}(a_k) = \text{a rnd elt from } \mathcal{T}(a_k)$
- $\mathcal{H}(a_i) = \mathcal{H}(a_{i+1})$  if it is in  $\mathcal{T}(a_i)$  else  $\mathcal{T}(a_i).peek()$

# Example run

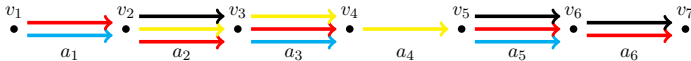


Figure: A path  $\mathcal{P}$

Start to compute  $\mathcal{T}(\mathcal{P})$

# Example run



Figure: Computing  $\mathcal{T}(\mathcal{P})$

$$\mathcal{T}(a_1) = \mathcal{F}(a_1)$$

# Example run

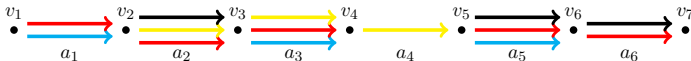


Figure: Computing  $\mathcal{T}(\mathcal{P})$

$$\mathcal{T}(a_2) = \mathcal{F}(a_2) \cap \mathcal{T}(a_1) \text{ since not empty}$$

# Example run

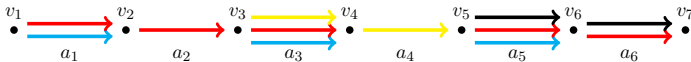


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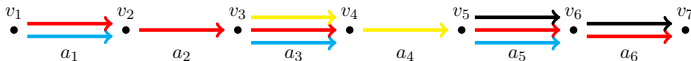


Figure: Computing  $\mathcal{T}(\mathcal{P})$

$$\mathcal{T}(a_3) = \mathcal{F}(a_3) \cap \mathcal{T}(a_2) \text{ since not empty}$$

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# Example run



Figure: Computing  $\mathcal{T}(\mathcal{P})$

$$\mathcal{T}(a_4) = \mathcal{F}(a_4) \text{ since } \mathcal{F}(a_4) \cap \mathcal{T}(a_3) = \emptyset$$

# Example run



Figure: Computing  $\mathcal{T}(\mathcal{P})$

$$\mathcal{T}(a_5) = \mathcal{F}(a_5) \text{ since } \mathcal{F}(a_5) \cap \mathcal{T}(a_4) = \emptyset$$

# Example run

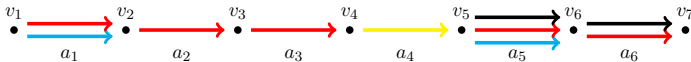


Figure: Computing  $\mathcal{T}(\mathcal{P})$

$$\mathcal{T}(a_6) = \mathcal{F}(a_6) \cap \mathcal{T}(a_5) \text{ since not empty}$$

# Example run



Figure: Computing  $\mathcal{T}(\mathcal{P})$

Start to compute  $\mathcal{H}(\mathcal{P})$

# Example run



Figure: Computing  $\mathcal{H}(\mathcal{P})$

$$\mathcal{H}(a_6) = \text{black}$$

# Example run



Figure: Computing  $\mathcal{H}(\mathcal{P})$

$$\mathcal{H}(a_6) = \text{black}$$

# Example run



Figure: Computing  $\mathcal{H}(\mathcal{P})$

$$\mathcal{H}(a_5) = \text{black} \text{ since } \text{black} \in \mathcal{T}(a_5)$$

# Example run



Figure: Computing  $\mathcal{H}(\mathcal{P})$

$$\mathcal{H}(a_5) = \text{black} \text{ since } \text{black} \in \mathcal{T}(a_5)$$



# Example run



Figure: Computing  $\mathcal{H}(\mathcal{P})$

Nothing to do for  $a_4, a_3$  and  $a_2$  since they only have 1 color

# Example run



Figure: Computing  $\mathcal{H}(\mathcal{P})$

$$\mathcal{H}(a_1) = \mathcal{H}(a_2) \text{ since } red \in \mathcal{T}(a_1)$$

# Example run



Figure: Minimum cost assignation

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$$w(\mathcal{P}) = 2$$

# Proof sketch

## Algo Part 1.

Induction proof on the length  $k$  of  $\mathcal{P}$ .



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Induction proof on the length  $k$  of  $\mathcal{P}$ .

If  $k = 1$  then  $w(\mathcal{P}) = 0$  which is optimal.



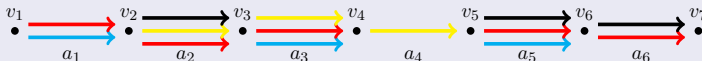
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Induction proof on the length  $k$  of  $\mathcal{P}$ .

We suppose the algo to be true for an arbitrary length  $k$ .

If  $\mathcal{F}(a_k) \cap \mathcal{F}(a_{k+1}) = \emptyset$



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Induction proof on the length  $k$  of  $\mathcal{P}$ .

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## Algo Part 1.

Induction proof on the length  $k$  of  $\mathcal{P}$ .

We suppose the algo to be true for an arbitrary length  $k$ .

If  $\mathcal{T}(a_k) \cap \mathcal{F}(a_{k+1}) = \emptyset$  and  $\mathcal{F}(a_k) \cap \mathcal{F}(a_{k+1}) \neq \emptyset$



# Proof sketch

## Algo Part 1.

Induction proof on the length  $k$  of  $\mathcal{P}$ .

Done



## Algo Part 2.

The number of CS inside  $\mathcal{T}$  is the same as the number of CS inside  $\mathcal{H}$ .



# Time Complexity

The algo is made by two sub-procedures:

---

Recall the first part:

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Complexity:

- First part :  $\mathcal{O}(k * |\mathcal{C}|)$

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Recall the second part:

- $\mathcal{H}(a_k)$  = a rnd elt from  $\mathcal{T}(a_k)$
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Complexity:

- First part :  $\mathcal{O}(k * |\mathcal{C}|)$
- Second part :  $\mathcal{O}(k * \log |\mathcal{C}|)$

# Time Complexity

Complexity:

- First part :  $\mathcal{O}(k * |\mathcal{C}|)$
- Second part :  $\mathcal{O}(k * \log |\mathcal{C}|)$

---

Global complexity:  $\mathcal{O}(k * |\mathcal{C}|)$ .

This complexity is optimal wrt the entry of the problem.

# Minimize CS in Graph

*Strategy:* Use the *MDD* data structure

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A state of a *MDD* is:

$\{\text{name: String, cost: Int, colors: Set of Colors}\}$

# Algorithm

- The root = {name:  $s$ , cost: 0, colors:  $\mathcal{C}$ }

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**Algorithm 1:** Construction of the layer  $\mathcal{L}_{i+1}$ 

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```
for  $\forall st \in \mathcal{L}_i$  do
  for  $\forall v \in \text{succ}(st.name)$  do
    col =  $\mathcal{F}(st.name, v)$ ;
    inter  $\leftarrow$  col  $\cap$  st.colors;
    if inter =  $\emptyset$  then
      |  $\mathcal{L}_{i+1}.add(\{name: v, cost: st.cost +$ 
      |   1, colors: col\})
    else
      |  $\mathcal{L}_{i+1}.add(\{name: v, cost: st.cost, colors: inter\})$ 
    end
  end
end
end
```

# MDD reduction

Let  $s_1$  and  $s_2$  two state on the same layer  $\mathcal{L}$ , having same name.

- *Dominated states*:  $s_1$  dominates  $s_2$  if the cost of  $s_1$  is smaller than the cost of  $s_2$
- *s-compatible states*:  $s_1$  and  $s_2$  are *s-compatible* if they have same cost. In this case,  $s_1$  and  $s_2$  are removed from  $\mathcal{L}$  and  $s_3 = \{\text{name: } s_1.\text{name}, \text{cost: } s_1.\text{cost}, \text{colors: } s_1.\text{colors} \cup s_2.\text{colors}\}$  is added to  $\mathcal{L}$



# Complexity

- Each layer at most has  $|V|$  states;
- The height of the *MDD* is  $k$ ;

The overall complexity is therefore

$$\mathcal{O}(k * |V|^2 * |\mathcal{C}|)$$

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Since  $|\mathcal{C}|$  is constant we can simplify and get:

$$\mathcal{O}(k * |V|^2)$$

# An example

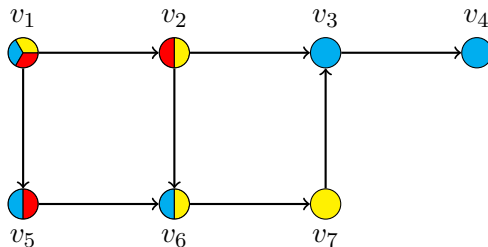
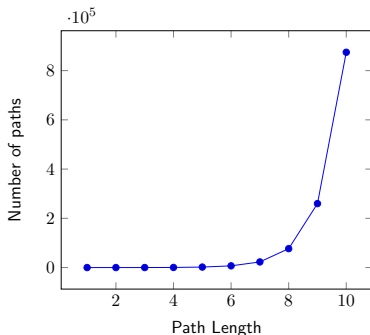


Figure: A colored graph example

Let's take  $k = 5$

# Benchmark: Solutions



(a) Classic algorithm

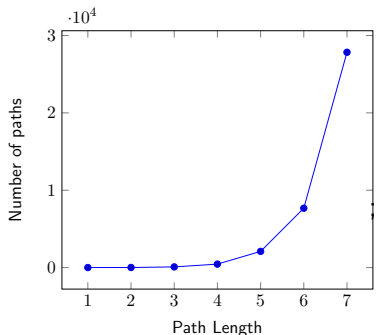
(b) *allDiff* constraint

Figure: Number of paths of a given length from the node 1

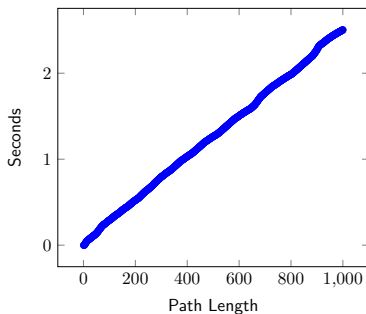
# The *allDiff* variant

A variant we can add to the problem is the introduction of the *allDiff* constraint to the nodes of the path.

- This modification entails:
  - Maintain a trace of the fathers of the current state,
  - Two states can be reduced only if they have same fathers.
- The second point causes a complexity blow up: the layer size can be  $2^{|V|}$ .
- The algorithm has now an exponential complexity.

## Benchmark

## Benchmark: Time



(a) Classic algorithm

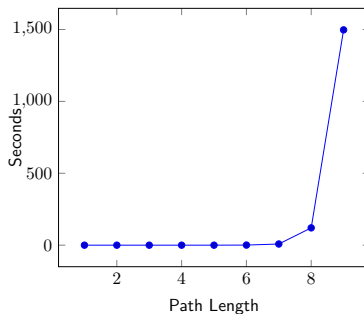
(b) *allDiff* constraint

Figure: Time taken to compute paths of a given length from the node 1

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*Thanks!*