

# Master Thesis – Prove Problem

Kefang Ding

January 15, 2019

---

## 1 Introduction

This article is used to explain the difficulty I confront about proving the soundness of generated model. In the first phase of algorithm, the existing model, positive event log and negative event log are used to generate a new directly-follows graph. Based on Inductiver Miner, this graph is transformed into a sound petri net without long-term dependency.

In the next phase, our algorithm focuses on detecting and adding long-term dependency in Petri net. We define, if the supported connection on the set pair of xor branches is over a threshold, pair has significant correlation. Therefore, this pair has long-term dependency.

During the implementation, it comes clear that supported connection only on the positive and negative event log is not enough, since the existing model can keep some directly-follows relations about xor branches which do not show in the positive event log or shows only in negative event log. Consequently, when we detect this long-term dependency on those xor branches, there is no evidence of long-term dependency on those xor branches. It results in an unsound model as shown in Fig , since those xor branches can't get fired to consume the tokens generated from the choices before.

## 2 Problem Description

To make the generated Petri net sound, we propose a method to incorporate the existing model on the long-term dependency detection. The definition ?? is rephrased into weight.

**Definition 2.1** (Rephrased Correlation of xor branch). *The correlation for two branches is expressed into*

$$Wlt(XORB_X, XORB_Y) = Wlt_{text}(XORB_X, XORB_Y) + Wlt_{pos}(XORB_X, XORB_Y) - Wlt_{neg}(XORB_X, XORB_Y)$$

, where  $Wlt_{text}(XORB_X, XORB_Y) = \frac{1}{\#XORB_Y}$ ,  $\#XORB_Y$  means the number of possible directly-follows xor branches  $XORB_Y$  after  $XORB_X$ .

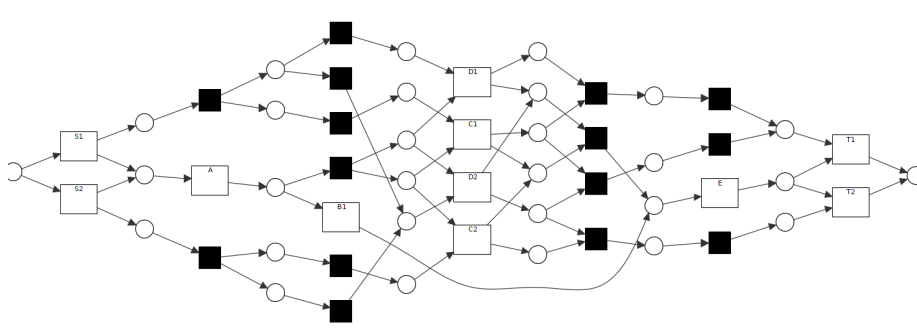


Figure 1: Unsound Repaired Model at Transition B1.

$$Wltpos(XORB_X, XORB_Y) = \frac{F_{pos}(XORB_X, XORB_Y)}{F_{pos}(XORB_X, *)},$$

$$Wltneg(XORB_X, XORB_Y) = \frac{F_{neg}(XORB_X, XORB_Y)}{F_{neg}(XORB_X, *)},$$

The  $F_{pos}(XORB_X, XORB_Y)$  and  $F_{neg}(XORB_X, XORB_Y)$  are the frequency of the coexistence of  $XORB_X$  and  $XORB_Y$ , respectively in positive and negative event log.

With this rephrased definition, to make the model sound, we need to prove, if there is a xor branch  $XORB_Y$  in the generated process tree, there must exist one long-term dependency related to it,  $\exists XORB_X, Wlt(XORB_X, XORB_Y) > lt - threshold$ . We formalize this problem. Else, the model can't be sound!!

**Theorem 2.1.** *Given a process tree, a pair of xor branch set,  $(B_A, B_B)$  with  $B_A = XORB_{X1}, XORB_{X2}, \dots, XORB_{Xm}$ ,  $B_B = XORB_{Y1}, XORB_{Y2}, \dots, XORB_{Yn}$ , the obligatory part between  $B_A$  and  $B_B$  is marked  $M$ , it is to prove:: for one xor branch  $XORB_Y$ , if  $W(M, XORB_Yj) > threshold$ , then there exists one  $XORB_Xi$  with*

$$Wlt(XORB_Xi, XORB_Yj) > lthreshold$$

Given a simplified scenes, it is listed in Fig 2, if there exists directly-follows relation of  $M$  and  $Y1$ , then there must exist one long-term dependency of  $Xi$  and  $Y1$ .

The definition of  $W(M, XORB_Yj)$  is reviewed below.

**Definition 2.2** (Assign new weights to graph  $G_{new}$ ). *there are three weights from  $G_{pos}$ ,  $G_{neg}$  and  $G_{ext}$ , the new weight is*

- For one directly-follows relation,

$$Weight(E_{G_{new}}(A, B)) = Weight(E_{G_{pos}}(A, B)) + Weight(E_{G_{ext}}(A, B)) - Weight(E_{G_{neg}}(A, B))$$

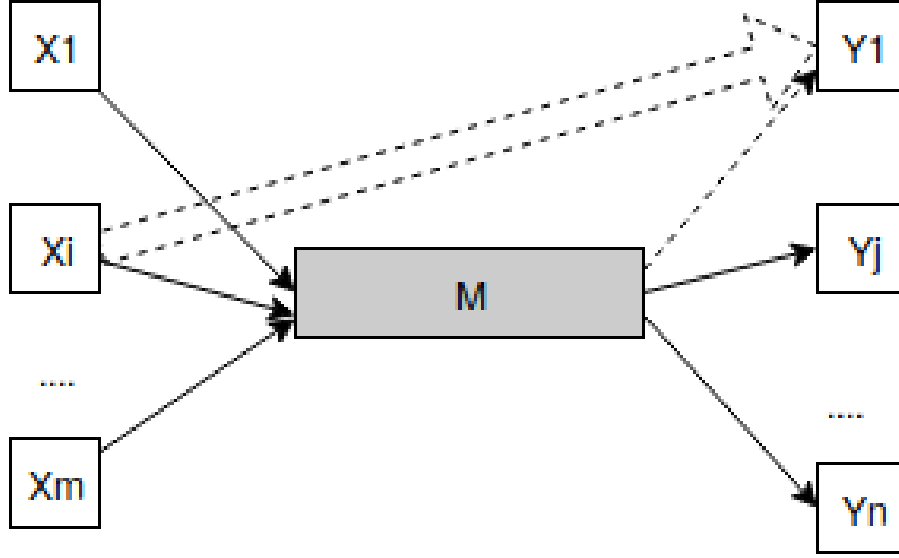


Figure 2: Unsound Repaired Model at Transition B1.

- Given a directly-follows graph  $G(L)$ , the weight of each directly-follows relation is defined as

$$Weight(E(A, B)) = \frac{Cardinality(E(A, B))}{Cardinality(E(A, *))}$$

When prove by contradiction, we assume that the opposite proposition is true. If it shows that such an assumption leads to a contradiction, then the original proposition is valid.

**Theorem 2.2.** *for one xor branch  $XORB_Y$ , if  $W(M, XORB_Y j) > threshold$ , there exists no one  $XORB_X i$  with*

$$Wlt(XORB_X i, XORB_Y j) < lt - threshold$$

Or we change to another thinking way to get the relation of threshold and lt-threshold, such that we have the theorem valid. Then we rephrase the question into

**Theorem 2.3** (Another way of thinking). *Given a process tree, a pair of xor branch set,  $(B_A, B_B)$  with  $B_A = XORB_{X1}, XORB_{X2}, \dots, XORB_{Xm}$ ,  $B_B = XORB_{Y1}, XORB_{Y2}, \dots, XORB_{Yn}$ , the obligatory part between  $B_A$  and  $B_B$  is marked  $M$ . If,*

*for one xor branch  $XORB_Y$ , if  $W(M, XORB_Y j) > threshold$ , there exists one  $XORB_X i$  with*

$$Wlt(XORB_X i, XORB_Y j) > lt - threshold$$

What is the relation of threshold and lt-threshold??

If we expand the theorem, we need to prove

**Theorem 2.4** (Relation of threshold and lt-threshold). *What is the relation of threshold and lt-threshold, to make the following proposition valid. If*

$$\begin{aligned}
& W(M, XOR_{Yj}) > threshold \\
& Weight(E_{G_{new}}(M, XOR_{Yj})) = Weight(E_{G_{pos}}(M, XOR_{Yj})) \\
& + Weight(E_{G_{ext}}(M, XOR_{Yj})) - Weight(E_{G_{neg}}(M, XOR_{Yj})) > threshold \\
& \frac{1}{|Y *|} + \frac{\sum_{Xi} Cardinality(M, Yj|Xi)}{\sum_{Xi} Cardinality(M, Y * |Xi)} - \frac{\sum_{Xi} Cardinality(M, Yj|Xi)'}{\sum_{Xi} Cardinality(M, Y * |Xi)'} \\
& \text{Then, exist one } Yj \text{ with} \\
& Wlt(Xi, Yj) > lt - threshold \\
& Wltext(Xi, Yj) + Wltpos(Xi, Yj) - Wltneg(Xi, Yj) > lt - threshold \\
& \frac{1}{|Y *|} + \frac{Cardinality(M, Yj|Xi)}{Cardinality(M, Y * |Xi)} - \frac{Cardinality(M, Yj|Xi)'}{Cardinality(M, Y * |Xi)'} > lt - threshold \\
& \text{Or there is a contradiction when all } Yj \\
& Wlt(Xi, Yj) < lt - threshold \\
& \sum_{Xi} Wlt(Xi, Yj) < |X *| lt - threshold \\
& \frac{|X *|}{|Y *|} + \sum_{Xi} \frac{Cardinality(M, Yj|Xi)}{Cardinality(M, Y * |Xi)} - \sum_{Xi} \frac{Cardinality(M, Yj|Xi)'}{Cardinality(M, Y * |Xi)'} < |X *| lt - threshold
\end{aligned}$$

$Cardinality(M, Yj|Xi)$  means the frequency of coexistence of  $M$  and  $Yj$  given  $Xi$  in the trace in positive, while  $Cardinality(M, Yj|Xi)'$  represents the frequency in negative.  $Cardinality(M, Y * |Xi)$  is the sum frequency of set  $Y1, ..Yj, ..Yn$ , it equals to

$$Cardinality(M, Y * |Xi) = \sum_{Yi} Cardinality(M, Yi|Xi)$$