

Master Thesis – Prove Problem

Kefang Ding

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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 $XORB_X$.

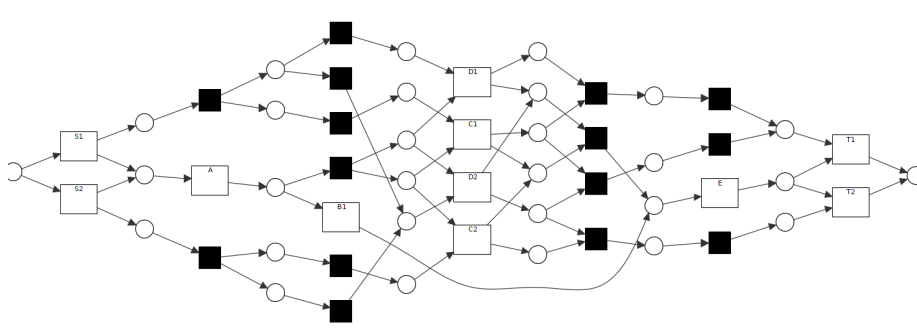


Figure 1: Unsound Repaired Model at Transition B1.

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

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

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!!

Proposition 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,

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

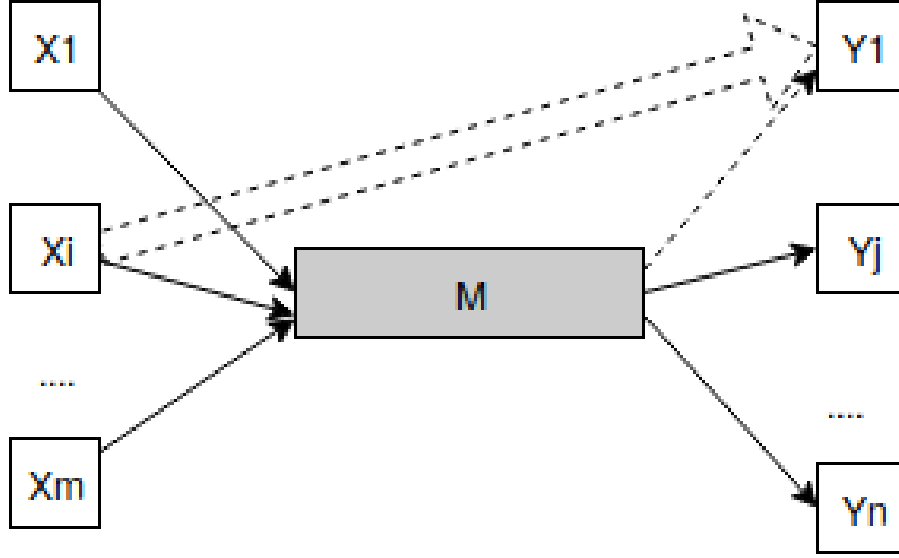


Figure 2: Unsound Repaired Model at Transition B1.

- 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.

Proposition 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

Proposition 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

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

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

$Cardinality(M, Yj|Xi)$ 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

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

If we set them into zero, there is a lot of existing edges kept into the old method with no evidence in event log to support the connection.