

# EpTO: An Epidemic Total Order Algorithm for Large-Scale Distributed Systems

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**Abstract**—One of the fundamental problems of distributed systems is that a message between two processes, may be delivered zero, or many times. This is way ordering events and messages is an essence in distrusted systems. There has been many algorithm to overcome this problem. EpTO is one of the recently introduced total order algorithms for large-scale distributed systems. EpTO is the first algorithm that claim to provide total order and scalability at the same time. We are going to verify this claim by implementing EpTO and evaluate it in a real-world condition.

## I. INTRODUCTION

Ordering of events is one the fundamental problems of distributed systems. Creating an algorithm which provide scalability, integrity and validity, along with an ordering for the event through all the peers and processes is one the hot topics in distributed systems research areas for many years. One of the recently designed algorithms on this topic is EpTO [1]. EpTO is an algorithm which is claimed to provide integrity, validity and total order throughout a distributed system. In addition, EpTO is designed to work without a global clock or synchronization process, that is why EpTO is well-suited for dynamic large-scale distributed systems.

There are many other algorithm for dissemination and ordering events in a distributed systems. There are some deterministic algorithms, which guarantee total order, agreement or other strong properties deterministically. But this type of algorithms are not scalable enough to be used in a large-scale distributed system [2, 3].

The problem with existing deterministic total ordering protocols, is that they need some sort of agreement between all peers in the system. This causes a massive amount of network traffic and overhead on the system. Moreover, an agreement feature for an asynchronous system requires to explicitly maintain a group and have access to a failure detector [4, 5]. Due to the faults and churn in large-scale distributed systems, failure detector turns into bottleneck of the system and thus, limits the scalability of the algorithm.

As an alternative for deterministic algorithms, there are probabilistic algorithms, focusing on scalability and resiliency against failures using a probabilistic dissemination approach [6, 7, 8, 9, 10, 11, 12, 13]. These algorithms guarantee to disseminate events through the system with high probability. In this way, there is not need for failure detectors and redundant traffic. Thus, these algorithms can be quit scalable. Although because these type of algorithms focus on reliability of dis-

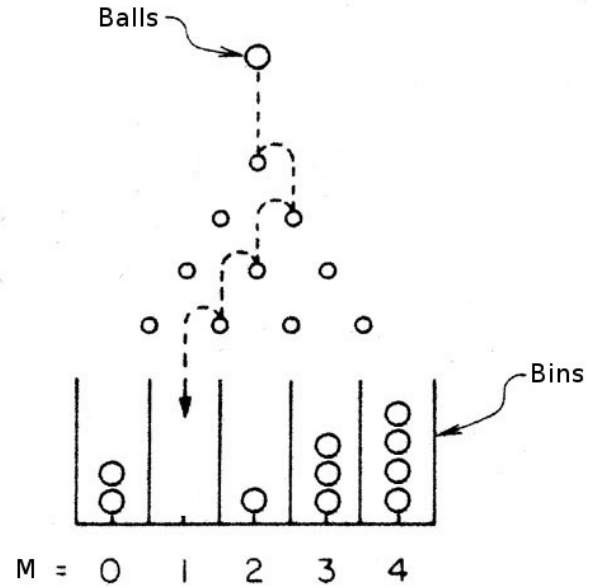


Figure 1. Balls-and-Bins

semination, they often ignore other needed properties such as ordering.

This is where EpTO is coming into the light. EpTO with mixing these two types of algorithms, provide total order along with scalability, validity and integrity. EpTO consists of two parts. First part is probabilistic dissemination. EpTO guarantees that all peers will receive an event with high probability. Second part of EpTO is deterministic ordering. Once peers receive all events, they will deterministically order them using the events timestamp, and in case of a tie using event's broadcaster's id.

For the first part, EpTO is using *balls-and-bins* approach [13]. The balls-and-bins problem is a basic probabilistic problem: consider  $n$  balls and  $m$  bins where we consequently throw balls into a bin completely random and independent from other balls. In this scenario, one the natural questions that comes to mind is that: what is the minimum number of balls that should be thrown, so that every bin has at least one ball?

Using the balls-and-bins approach we model processes as bins and events as balls. Based on this model we will calculate how many balls need to be *thrown* such that every bin contains at least one ball with arbitrarily high probability. Using this

approach the number of messages transmitted per process per round is logarithmic in the number of processes, therefore the number of messages sent in the system is low and uniform over all processes. Using these approaches, EpTO becomes highly scalable and resilient, having in mind that it also provides total order.

#### A. Contribution

Until now, the creators of EpTO test this algorithm in a simulated environment, and not in a real-world scenario. We are going implement EpTO in pure JAVA using NeEM library [14]. Afterwards we will evaluate EpTO by comparing it to the deterministic total order algorithm provided by JGroups [15] in both stable (no-churn) and unstable systems. This comparison help us to verify if EpTO is actually perform as expected in real-world conditions or not.

These tasks will be split into two parts. The first part will assume that all peers are known by all peers. We will implement EpTO using this assumption and test it against JGroups. I the second part of the tasks, will assume unknown membership on all peers. We will update EpTO and use a PSS<sup>1</sup>. We will also update the underlying layer NeEM to use UDP instead of TCP and switch to NIO.2<sup>2</sup>.

### II. ORDERING ALGORITHMS

We will explain why Ordering algorithms are necessary, the difference between partial order and Total order and why partial order might not be sufficient in some cases

Distributed systems, like centralized systems need to preserve the temporal order of event produced by concurrent process in the system. When there are separated processes that can only communicate through messages, you can't easily order these messages. Therefore we need ordering algorithms to overcome this problem.

We have two types of ordering algorithms [3]: the partial order algorithms and the total order algorithms.

#### A. Partial Order Algorithms

Assuming  $S$  is partially ordered under  $\leq$ , then the following statements hold for all  $a, b$  and  $c$  in  $S$ :

- Reflexivity:  $a \leq a$  for all  $a \in S$ .
- Antisymmetry:  $a \leq b$  and  $b \leq a$  implies  $a = b$ .
- Transitivity:  $a \leq b$  and  $b \leq c$  implies  $a \leq c$ .

#### B. Total Order Algorithms

A totally ordered set of events is a partially ordered set which satisfies one additional property:

- Totality (trichotomy law): For any  $a, b \in S$ , either  $a \leq b$  or  $b \leq a$ .

Explain advantages and usefulness of both systems

In other words, total order is an ordering that defines the exact order of every event in the system. On the other hand, partial ordering only defines the order between certain key

<sup>1</sup>Peer Sampling Service

<sup>2</sup>Non-blocking I/O available from Java 7

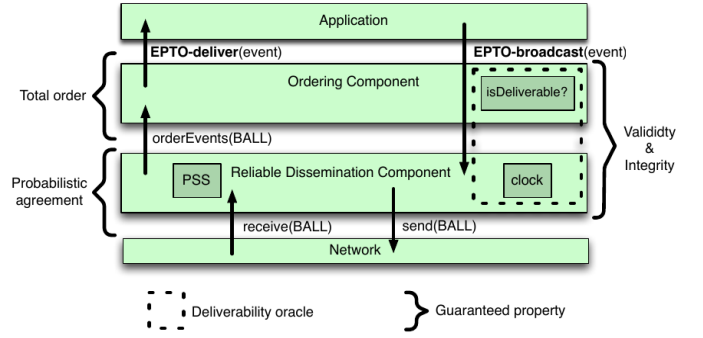


Figure 2.

We need a better quality figure (SVG)

4.

events that depend on each other. Partial order can be useful since it is less costly to implement but, in some cases where the order of all events is important, for example when we need an exact image of the state of the system at all time, we have to use total order.

### III. EpTO

#### A. Definitions

A process or peer is defined as an actor in our system running the application that needs total order. Each process will communicate with other processes in the distributed systems and exchange events and order them together.

An event is defined as something that happens at a given time. For example, we could imagine a system where each process publish some sort of data. The moment where this data is published combined with the data is called an event.

A ball is a set of events bundled together and sent as one package. We use balls in EpTO to reduce network traffic and make it scalable in terms of processes and events.

EpTO is scalable in the sense that

Explain the logarithmic scalability

Since EpTO is using a probabilistic agreement instead of a deterministic agreement, there might be a situation where a peer does not receive an event (with a very low arbitrary probability). In this case there will be a hole in the sequence of delivered events but even in this case, the order of the delivered event will be protected by EpTO's deterministic ordering algorithm and total order property is preserved.

#### B. Dissemination Component

The Dissemination component is the component that bridges EpTO with the rest of the network. As we can see in Figure 2, it is in charge of receiving balls, open them, pass them to the Ordering component and then forward them to  $K$  other processes, where  $K$  denotes the gossip fan-out.

When an application wants to publish an event, it will broadcast this event to the Dissemination component.

### C. Ordering Component

The Ordering component is responsible for ordering events before delivering them to the application. To achieve this, the Ordering component has a *received* hash table of (*id*, *event*) pairs containing all the events which are received by the peer, but not yet delivered to the application and a *delivered* hash table containing all the events which are delivered.

In brief, the Ordering component first increments the timestamp of all the events which have been received in previous rounds to indicate the start of a new round. Then, it processes new events in the received ball by discarding events that have been received already (delivered events or events with timestamp smaller than the last delivered event). This is done to prevent delivering duplicate events. The remaining events in the received ball will be added to the *received* hash table, and wait to be delivered based on the Stability oracle.

### D. Stability Oracle

The Stability oracle is the component that outputs timestamps. It will increment its local clock every round as well as synchronize itself using timestamps of newly received events, to make sure the local clock does not drift too much.

This Stability oracle offers an API to the Ordering component letting it know when an event is mature enough to be delivered to the application.

As this component is local and only correct itself when we receive new events, it generates no network traffic. This means it does not impact the scalability of EpTO.

## IV. PERFORMANCE COMPARISONS

Here we will present the different test we ran, the methodology used and the result we obtained

### A. Methodology

Here we explain how we ran our tests and on what type of machines

### B. Peer membership known

We will present the results obtained and try to explain the results

### C. Peer membership unknown

We will present the results obtained and try to explain the results

## V. CONCLUSION

He we will summarize the results we found and present some future tasks that could be accomplished on EpTO

## ACKNOWLEDGMENT

Write acknowledgment

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