

A journey into PostgreSQL logical replication

The Next Chapter

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The journey: prelude

Toggl Track evolved on top of a monolithic transactional PostgreSQL database.

With an ever-increasing dataset, we started to feel the need to move away from the transactional normalized - data structure to provide the analytical features that users were asking for.

The journey: the plan

Move it, transform it. Use it.

Leverage logical replication as a means of facilitating close to real-time transformation of our transactional data. Make it the source of a CDC pipeline.

We implemented our own logical replication client in GoLang, using **pglogrepl** *(github.com/jackc/pglogrepl)***.**

The journey: the plan

A major product requirement for our analytic features was near **real-time** availability.

Logical replication was our best bet to achieve it. And while at it, we managed to provide the means to "liberalize" access to "data changes".

Later on used to decrease latency in other processes such as internal BI tools, email notifications, or simply data propagation from sources of truth to other services.

The journey: the bad turn

Caring for data-changing events - only - turned out to be a **bad turn**.

Logical replication streams data that is already committed.

It won't be rolled back. And **DDL changes will not be there anyway**.

The journey: consumption client

While developing our consumption client, our line of thinking was: **no use keeping track of events that don't change data**.

Like transaction begins and commits.

We cared only for Inserts, updates, deletes, and maybe truncates.

Spoiler! It was a crappy idea.

Developing our Consumption Client

Logical replication slot

The slot is persistent, regardless of an active connection. And will store the consumption status, using two offsets, restart, and flush LSN*.

LSNs are pointers to given locations in the WAL. Logical replication clients must periodically push consumption status to update the slot.

* https://www.postgresql.org/docs/current/datatype-pg-lsn.html

LSN Examples

BEGIN 4/98EE65C0 INSERT 4/98EE65C0 UPDATE 4/98EE66D8 UPDATE 4/98EE6788 COMMIT 4/98EE6830

START TRANSACTION;

INSERT INTO track (description, duration) VALUES ('Reading', 360000); UPDATE track total SET duration = duration + 360000; UPDATE user SET entries = entries + 1; COMMIT;

BEGIN 4/98EE6950 UPDATE 4/98EE6AD8 UPDATE 4/98EE6D28 UPDATE 4/98EE6DD8 COMMIT 4/98EE6F30

START TRANSACTION; UPDATE track SET duration = duration + 360000; UPDATE track total SET duration = duration + 360000 ; UPDATE users SET entries = entries + 1; COMMIT;

BEGIN 4/98EE6F68 INSERT 4/98EE6F68 COMMIT 4/98EE7040

INSERT INTO users (email, password) VALUES ('@.com', '...');

Without Concurrency

Without Concurrency

WAL

With Concurrency

With Concurrency

Concurrency

As we were intentionally disregarding transactions, all we had to work with were data-changing events and their offsets.

Fast forwarding…

We attempted to live with the conditions that we found outside dev environments.

A mix of **bad assumptions**, the **ideal conditions for confusion** due to our first use case *(summing up tracked time / commutative property)*, and **lack of knowledge** about logical replication inner works led to dark times trying to figure out why our brand new OLAP data was - sometimes - inconsistent.

Bad: Assuming incremental LSNs

We assumed that LSNs were incremental cross-transactions.

Every logical replication event comes with an LSN offset which corresponds to a location in the WAL, but logging happens concurrently.

Bad: Assuming incremental LSNs

We first attempted to solve the issue by tracking the current LSN and **discarding data with offsets smaller** than our current position. We were under the wrong impression that we would have incremental LSN offsets.

In this example by filtering for L NS $>$ "5" we would discard the first data event from the next transaction.

Bad: Commiting ops offsets

After figuring out that our data events presented "out-of-order" offsets. We fell into another nuance when we stopped discarding data based on expecting an incremental offset:

We would duplicate it on reconnection, by committing operation offsets.

Key points

A few key points became clear laying the path to a proper implementation:

Key points

- \vee The replication stream data is sorted data by transaction end offsets
- \vee When we commit LSN offsets that pertain to mid-transaction events, pg **will resent the whole transaction again** upon reconnection.
- \vee Events for a given transaction are always streamed together, regardless of log positioning.

BEGIN **LSN000001** INSERT **LSN000002** UPDATE **LSN000005** COMMIT **LSN000006** BEGIN **LSN000003** UPDATE **LSN000004** UPDATE **LSN000007** COMMIT **LSN000008**

Key points

BEGIN **LSN000001** INSERT **LSN000002** UPDATE **LSN000005** COMMIT **LSN000006 T1**

Committing offset "6ˮ doesn't prevent data in the next transaction from being sent.

BEGIN **LSN000003** UPDATE **LSN000004** UPDATE **LSN000007** COMMIT **LSN000008 T2**

All transaction events are resent upon reconnection if the offset that we **committed to the replication slot is not the transaction end** - or bigger.

Conclusion

By always committing the transaction end LSN we make sure that we process all data-changing events respecting their transactional integrity.

Only **committing the appropriated offset will allow us to correctly manage the slot state**. *That is, without additional stateful logic on the client side.*

The Next Chapter

Our vision

The CDC pipeline created by our logical replication client became the source of data for many of our endeavors:

- Generalist **data aggregations**.
- User-generated reporting.
- Real-time business intelligence reporting.
- **Our authorization system.**

Our vision

On top of our CDC pipeline, we can then achieve whatever transformations are needed.

Scale horizontally and, if we choose to, segregate by feature or general domain.

How it's going: domain segregation

To better scale, we also started **decoupling domains from our main service**.

As a result, our **authorization system** needed to **source data from different DBs**:

- Authorization *(roles, permissions, and their relations to users)*.
- Subscriptions *(plans, features, status)*.
- Remaining transactional data *(subject to business logic)*.

How it's going

As we split domain responsibilities into different services, more sources of data were added.

How it's going: domain segregation

Dealing with logical replication inner works was just the beginning of the journey.

Authorization use-case

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Our freshly developed CDC couldn't look better to achieve this goal.

We figured that we could simply ingest data from all three sources and pre-compute user session data.

We don't want users to have to wait very long for their 50k project list to load.

Authorization use-case

Pre-computed Sessions

Will tell us what workspaces a user has access to, with which permissions, and the projects that he can see in them.

Behind the scenes, our **OLAP transformation is pre-generating user sessions** by applying subscription, authorization, and business logic to data changes.

Authorization use-case

Sessions are expected to be **updated fast**. Users won't want to wait very long for their newly created project to be available for use.

That's **challenging**.

Especially if you have users with 100k project lists.

Tips & Tricks

Cross our implementations, while striving for near "real-time" data transformations, we collected a few do's and don't do's.

Tips & Tricks: Embrace duplication

While the proper use of LSN event offsets **fully prevents data loss, it doesn't prevent data duplication**. On plug-off events, it will happen.

Embrace that possibility. Take advantage of event atomicity on the OLAP transformation and make sure that the operations are idempotent.

Tip?

Use of an **updated** timestamp in your transactional tables. Make sure that every change correctly updates it *(ie.: through triggers)*. And make use of it in your transformations to dedup.

Tips & Tricks: Bulk updates

Processing data changes individually is challenging.

Prepare your pipeline to deal with events in bulk.

Tip?

Add a small delay in your data collection and bulk upload changes. **Ingest and transform data in bulk**. *Use upsert to keep it a single operation.*

Tips & Tricks: Bulk updates

INSERT INTO transformation (

```
 user_id,
total time,
 updated_at
```
SELECT

```
 (d->>'user_id')::int AS user_id,
   (d->>'total_time')::int AS total_time,
   (d->>'updated_at')::timestamp without time zone AS updated_at
FROM JSON ARRAY ELEMENTS ($1::jsonb) d
ON CONFLICT ON CONSTRAINT transformation pkey
DO UPDATE SET
  total time = excluded.total time,
  updated at = excluded.updated at
WHERE
   transformation.updated_at <
    excluded.updated at;
```


Tips & Tricks: Statement triggers

If we update multiple rows, the transformation will run multiple times with **row-based triggers**. Make sure that logic uses statement changes instead.

Tip?

Use **FOR EACH STATEMENT** triggers to the detriment of **FOR EACH ROW** triggers. If you must use row triggers, use conditioning to make sure that the transformation is only executed if the change is relevant to it.

Tips & Tricks: Statement triggers

CREATE TRIGGER after_update **AFTER UPDATE ON** transformation **REFERENCING NEW TABLE AS** new_table **OLD TABLE AS** old_table FOR EACH STATEMENT EXECUTE FUNCTION olap.update team goals();

CREATE OR REPLACE FUNCTION olap.update_team_goals() RETURNS trigger LANGUAGE plpgsql **AS \$\$**

BEGIN

```
 WITH changes AS (
     SELECT t.team_id, SUM(new_table.total_time - COALESCE(old_table.total_time, 0 )) AS total_time
     FROM relations.user teams t
     JOIN new table ON t.user id = new table.user id
     LEFT JOIN old_table ON new_table.user_id = old_table.user_id
     GROUP BY t.team_id
 )
    UPDATE olap.team_goals tg SET total_time = total_time + changes.total_time
    FROM changes WHERE tg.team_id = changes.team_id;
    RETURN NULL;
  END;
```
\$\$;

Tips & Tricks: Inclusive Indexes

When creating OLAP structures from normalized data changes, you will find yourself looking into relationship tables to find parent ids*, for instance*.

Using **inclusive indexes saves an extra trip to the table** to retrieve the information itself.

Tips & Tricks: Inclusive Indexes

```
CREATE SCHEMA relations;
CREATE TABLE relations.user_teams (
     user_id integer NOT NULL,
     team_id integer NOT NULL,
     CONSTRAINT user_teams_pkey PRIMARY KEY (user_id, team_id )
);
```

```
INSERT INTO relations.user_teams (user_id, team_id )
SELECT (random() * 10000)::int AS user id, (random() * 1000)::int AS team id
FROM generate series(1, 1000000)
GROUP BY 1, 2;
```
SELECT user_id, **COUNT(**1**) FROM** relations.user_teams **GROUP BY** user_id **ORDER BY** 2 **DESC LIMIT** 1**;**

CREATE INDEX simple_user_index **ON** relations.user_teams **(**user_id**); EXPLAIN (ANALYSE, VERBOSE) SELECT** team_id **FROM** relations.user_teams **WHERE** user_id = 1560 **;**

CREATE INDEX inclusive_user_index **ON** relations.user_teams **(**user_id**) INCLUDE (**team_id**); EXPLAIN (ANALYSE, VERBOSE) SELECT** team_id **FROM** relations.user_teams **WHERE** user_id = 1560 **;**

Tips & Tricks: Inclusive Indexes

EXPLAIN (ANALYSE, VERBOSE) SELECT team_id **FROM** relations.user_teams **WHERE** user_id = 1560

```
Index Scan using simple user index on relations.user teams (cost=0.42..4.29 rows=95 width=4)
(actual time=12.193..12.220 rows=128 loops=1)
  Output: team_id
  Index Cond: (user_teams.user_id = 1560)
Planning Time: 0.116 ms
Execution Time: 12.253 ms
```
vs

```
Index Only Scan using inclusive user index on relations.user teams (cost=0.42..3.19 rows=95
width=4) (actual time=0.077..0.089 rows=128 loops=1)
  Output: team_id
  Index Cond: (user teams.user id = 1560)
   Heap Fetches: 0
Planning Time: 0.129 ms
Execution Time: 0.112 ms
```


In a situation where **sourced data is unstructured**, we may want to expose some properties, so they are directly indexable, for instance.

Make use of generated columns to avoid scanning through unstructured data at run time.

CREATE SCHEMA olap; **CREATE TABLE** olap.unstructured **(** dump **jsonb NOT NULL)**;

```
INSERT INTO olap.unstructured (dump)
SELECT JSONB_BUILD_OBJECT('user_id', (random() * 10000)::int, 'total_time', (random() * 1000)::int, 
'goal', (random() * 1000) :: int + 1) FROM generate series(1, 100000000);
```
CREATE INDEX unstructured_gin_index **ON** olap.unstructured **USING GIN (**dump**)**;

EXPLAIN (ANALYSE, VERBOSE) SELECT COUNT(1**) FROM** olap.unstructured **WHERE** $(\text{dump->>}'\text{total time}')::int \geq (\text{dump->>}'\text{goal}')::int;$

ALTER TABLE olap.unstructured **ADD COLUMN** goal_completed **boolean GENERATED ALWAYS AS (**(dump->>'total_time')::int >= (dump->>'goal')::int **) STORED**; **CREATE INDEX** goal_completed_index **ON** olap.unstructured **(**goal_completed **)**;

EXPLAIN (ANALYSE, VERBOSE) SELECT COUNT(1**) FROM** olap.unstructured **WHERE** goal_completed;

EXPLAIN (ANALYSE, VERBOSE) SELECT COUNT(1**) FROM** olap.unstructured **WHERE** $(\text{dump->>}!\text{total time'})::\text{int} \geq (\text{dump->>}!\text{qoal'})::\text{int};$

Finalize Aggregate (cost=2292731.42..2292731.43 rows=1 width=8) (actual time=74036.280..74047.599 rows=1 loops=1)

Output: count(1)

 -> Gather (cost=2292731.00..2292731.41 rows=4 width=8) (actual time=74036.072..74047.584 rows=5 loops=1)

```
Output: (PARTIAL count(1))...
```
 -> Partial Aggregate (cost=2291731.00..2291731.01 rows=1 width=8) (actual time=74029.976..74029.977 rows=1 loops=5)

Output: PARTIAL count(1)...

 -> **Parallel Seq Scan** on olap.unstructured (cost=0.00..2270481.00 rows=8500000 width=0) (actual time=0.477..73068.822 rows=10190650 loops=5)

Filter: (((unstructured.dump ->> 'total time'::text))::integer >= **((unstructured.dump ->> 'goal'::text))::integer)** Rows Removed by Filter: 10209350... Planning Time: **10.810 ms**

Execution Time: **74048.038 ms**

EXPLAIN (ANALYSE, VERBOSE) SELECT COUNT(1**) FROM** olap.unstructured **WHERE** goal_completed;

—---

Finalize Aggregate (cost=692971.65..692971.66 rows=1 width=8) (actual time=3173.567..3195.634 rows=1 loops=1)

Output: count(1)

 $\left[-\right]$ Gather (cost=692971.23..692971.64 rows=4 width=8) (actual time=3173.488..3195.625 rows=5 loops=1)

Output: (PARTIAL count(1))...

 -> Partial Aggregate (cost=691971.23..691971.24 rows=1 width=8) (actual time=3165.396..3165.397 rows=1 loops=5)

Output: PARTIAL count(1)...

 -> Parallel **Index Only Scan** using goal_completed_index on olap.unstructured (cost=0.57..660276.85 rows=12677752 width=0) (actual time=0.161..2484.704 rows=10190650 loops=5) Output: goal_completed Index Cond: (unstructured.goal_completed = true) Heap Fetches: 0... Planning Time: **0.234 ms**

Execution Time: **3195.690 ms**

Tips & Tricks: JSONB

When handling or storing JSON, the binary format can offer advantages, but using **JSONB functions** for smaller - looping - sub-operations **degrades performance** due to the parsing overhead.

Execute operations in JSON, and cast the result to JSONB instead.

Tips & Tricks: JSONB

```
EXPLAIN (ANALYSE, VERBOSE) SELECT JSONB_BUILD_OBJECT('obj', obj) FROM (
     SELECT JSONB_OBJECT_AGG(user_id, tracked_obj) AS obj FROM (
          SELECT user_id, JSONB_OBJECT_AGG(team_id, JSONB_BUILD_OBJECT('total_time', (random() *
1000)::int)) AS tracked_obj FROM (
                SELECT (random() * 10000)::int AS user_id, (random() * 10000)::int AS team_id
                FROM generate series(1, 1000000)
          ) gen
          GROUP BY user_id
     ) agg
) foo;
Subquery Scan on foo (cost=50005.01..50005.03 rows=1 width=32) (actual time=6325.727..6368.489 
rows=1 loops=1)
  Output: jsonb_build_object('obj', foo.obj)
   -> Aggregate (cost=50005.01..50005.02 rows=1 width=32) (actual time=5503.496..5503.499 rows=1 
loops=1)
         Output: jsonb_object_agg((..., (jsonb_object_agg(..., jsonb_build_object('total_time',...)
        -> HashAggregate (... (actual time=3535.802..4119.971 rows=10001 loops=1)...
Planning Time: 0.174 ms
Execution Time: 6412.477 ms
```


Tips & Tricks: JSONB

```
EXPLAIN (ANALYSE, VERBOSE) SELECT JSONB_BUILD_OBJECT('obj', obj) FROM (
     SELECT JSON_OBJECT_AGG(user_id, tracked_obj) AS obj FROM (
          SELECT user_id, JSON_OBJECT_AGG(team_id, JSON_BUILD_OBJECT('total_time', (random() * 
1000)::int)) AS tracked_obj FROM (
                SELECT (random() * 10000)::int AS user_id, (random() * 10000)::int AS team_id
                FROM generate series(1, 1000000)
          ) gen
          GROUP BY user_id
     ) agg
) foo;
Subquery Scan on foo (cost=50005.01..50005.03 rows=1 width=32) (actual time=3295.757..3345.740 
rows=1 loops=1)
  Output: jsonb_build_object('obj', foo.obj)
   -> Aggregate (cost=50005.01..50005.02 rows=1 width=32) (actual time=1846.187..1846.190 rows=1 
loops=1)
         Output: json_object_agg((..., (json_object_agg(..., json_build_object('total_time',...)
        -> HashAggregate (... (actual time=1765.143..1776.300 rows=10001 loops=1)...
Planning Time: 0.104 ms
Execution Time: 3355.383 ms
```
Thank you.

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