

The logo for TigerBeetle, featuring a stylized 'T' icon followed by the text 'TigerBeetle' in a bold, sans-serif font.

**TigerBeetle**

# 1000X WORLD TOUR

13 CITIES | 6 DAYS





# TigerFans: High-Performance Ticketing with TigerBeetle

From Double-Entry Accounting to 977 Tickets per Second

★ TigerFans

TigerBench GitHub

## TigerFans 2025

A fictional conference for people who love fast, correct systems.  
Dec 3–4, 2025

[Order now](#)[Why attend?](#)

powered by

**TigerBeetle —  
1000x FASTER  
OLTP**

tigerbeetle cluster size

**1 (single-replica)**

reservations store

**Redis**

orders database

**PostgreSQL**

All transactions in the demo are mock-paid, inventory tracked with TigerBeetle.



It began with a tweet.

*"How would you sell Oasis-scale tickets without overselling?"*

Answer from **Joran**: Too easy: TigerBeetle.

I wanted to know:

- **HOW** do you actually model ticketing as **double-entry accounting**?
- What does the **real code** look like?
- How do you handle **timeouts, webhooks, idempotency**?

So I decided to **build the whole thing**.



## Goal: Realistic, Not Toy



I didn't want a toy example.

I wanted a realistic ticketing flow

- Checkout with multiple ticket classes & goodies
- Pending reservations with timeouts
- Webhook-based payment confirmation
- Strict no overselling guarantee

Stack:

- FastAPI (Python, async)
- TigerBeetle (accounting)
- SQLite at first
- MockPay (fake payment provider)



TigerBeetle gives us **accounts, transfers, debits, credits**.

We turn ticketing into an **accounting problem**:

- Every reservation is a **transfer**
- Every ticket class is a set of **accounts**
- **Debits = Credits** => system never goes out of balance

Double-entry accounting gives us:

- Built-in **error detection**
- **Auditability** for free
- And with TB: **durability + performance** in the same model



# Three Accounts per Ticket Class



For each ticket class:

- **Operator:**  
Holds all inventory
- **Budget:**  
What we're allowed to sell
- **Spent:**  
Tickets actually reserved/used

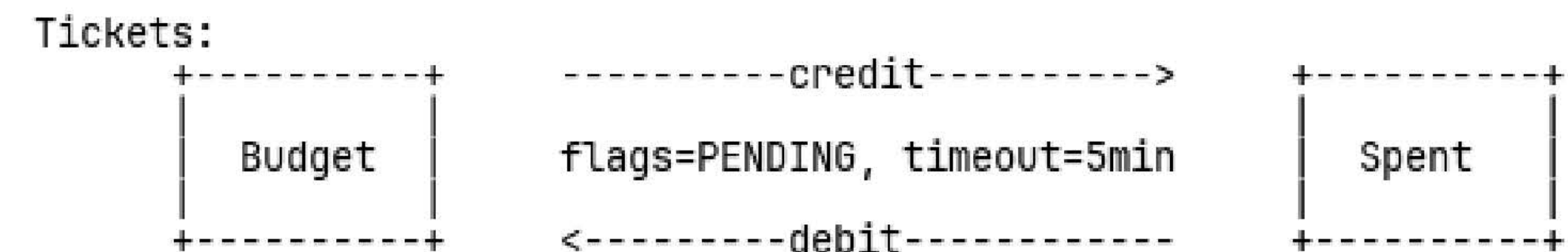
Flow:

1. Fund Budget from Operator
2. Move from Budget Spent on reservation
3. Post or void based on payment outcome

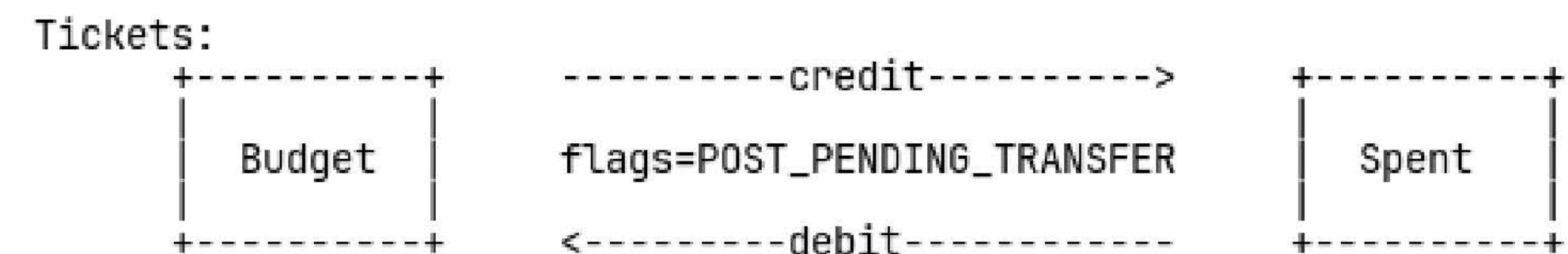
INIT FLOW: Fund the budget  
=====



HOLDING FLOW: Limited Hold on the resource  
=====



COMMIT FLOW: Post the pending transfers  
=====



## Pending Transfers & Timeouts



Checkout = **create pending transfer:**

- Budget -> Spent
- **5-minute timeout**
- **DEBITS MUST NOT EXCEED CREDITS** flag on accounts

Payment succeeds:

- Post the pending transfer => ticket is **locked in**

Payment fails / user disappears:

- Void the transfer => it just **vanishes**

No cron jobs. No cleanup races.

**Correctness is enforced by TigerBeetle's invariants.**



# Initial Architecture: Simple & Clean

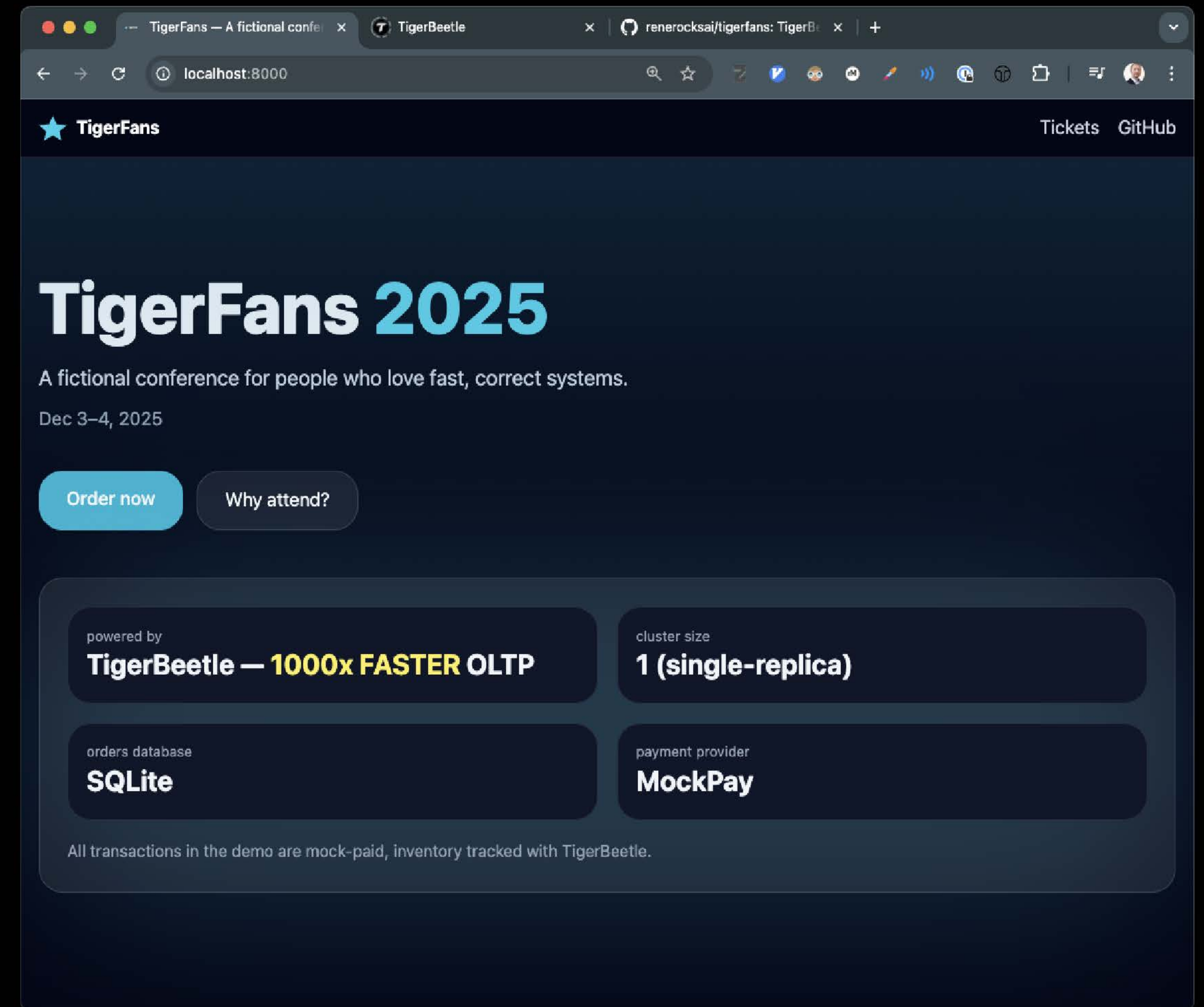


## Version 1:

- FastAPI (1 worker)
- SQLite
- TigerBeetle (dev mode)
- MockPay

## Everything worked:

- 2-phase checkout
- Timeouts
- Webhooks
- UI + Admin view







First measurement:

- **~115 tickets/sec**

Oasis baseline:

- ~1.4M tickets / 6 hours
- **≈ 65 tickets/sec**

So we're at about **O(1.7 Oasis)**.

I send this to Joran...





First measurement:

- **~115 tickets/sec**

Oasis baseline:

- ~1.4M tickets / 6 hours
- **≈ 65 tickets/sec**

...he replies:

*“I was surprised the TPS is so low.*

So we’re at about **O(1.7 Oasis)**.

*It should be ~10k.”*

I send this to Joran...



# Where Is the Time Going?

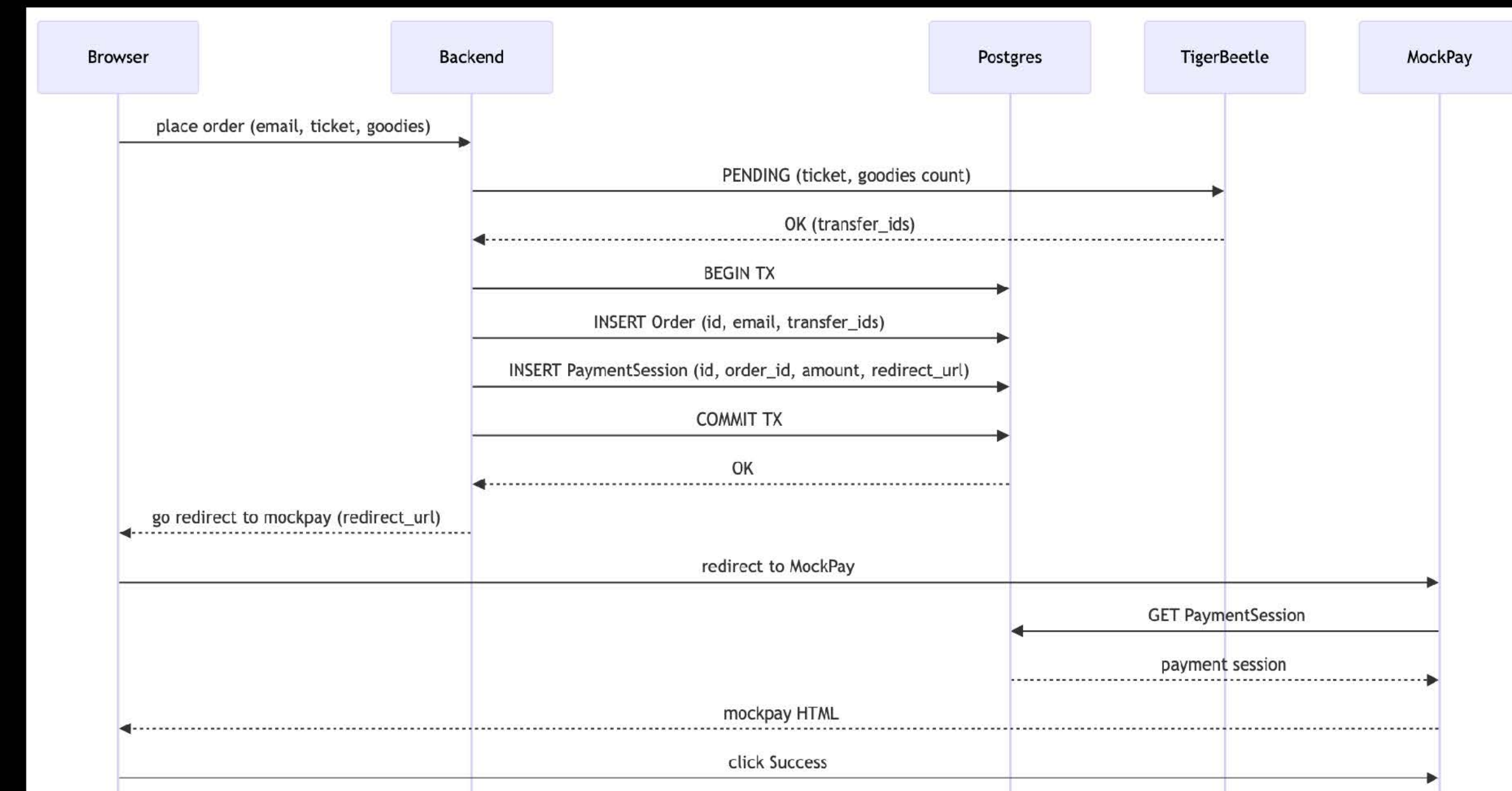


## Checkout flow:

1. Browser -> Server: *place order*
2. Server -> TB: PENDING transfers
3. Server -> DB: sessions + orders
4. Server -> Browser: redirect

## Webhook flow:

1. MockPay -> Server
2. Server -> DB: idempotency checks
3. Server -> TB: POST transfers
4. Server -> DB: final order write





# Bottleneck #1: PostgreSQL Everywhere



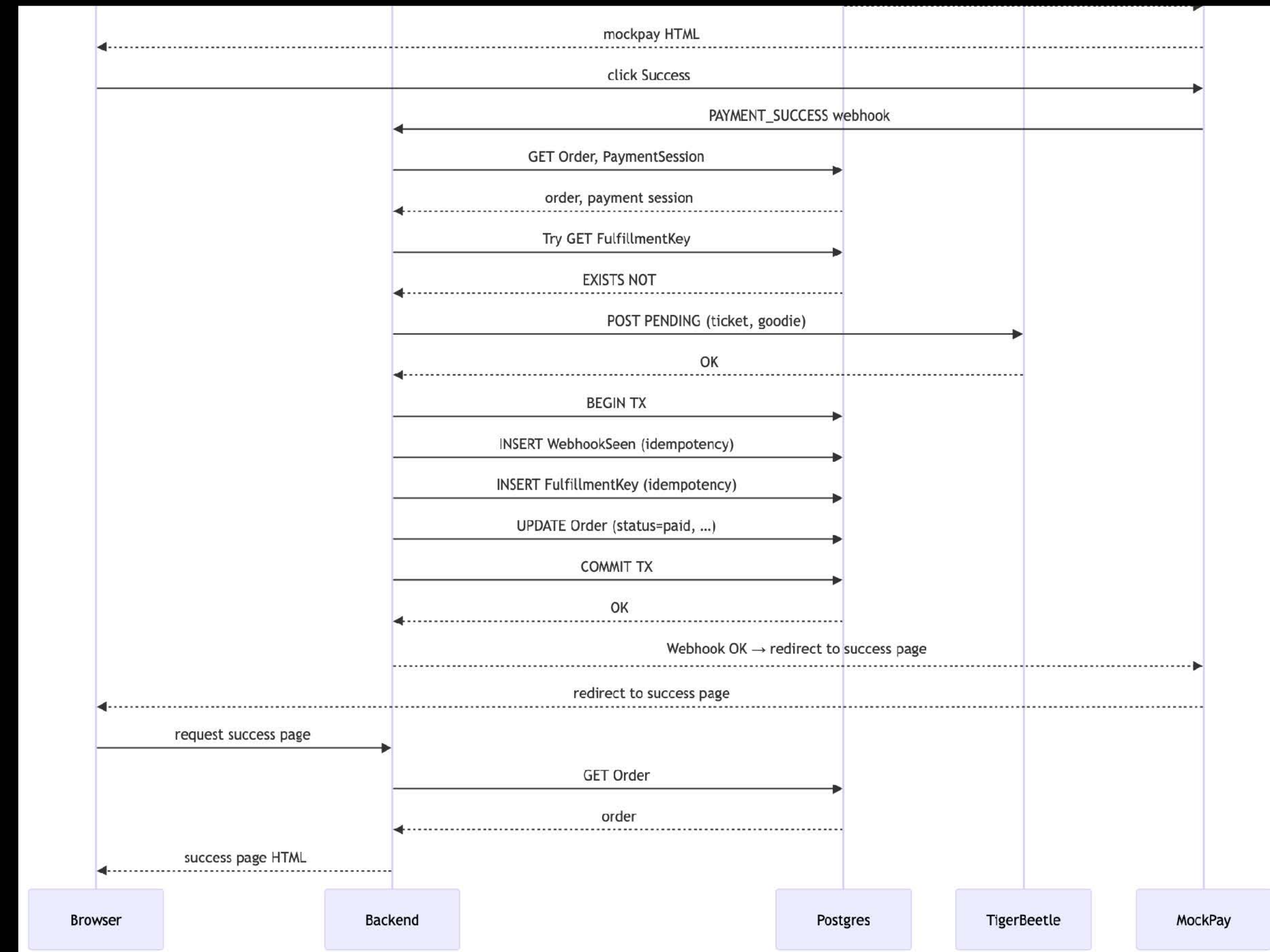
With PostgreSQL in the hot path:

- Reservations: **~150 ops/s**
- Webhooks: **~130 ops/s**

Every request:

- 2–4 round-trips to PG
- Multiple fsyncs
- Contention on the same tables

The database, not TigerBeetle, is the bottleneck.





## Redis Experiment: All-In Memory



I swap PostgreSQL -> Redis for everything :

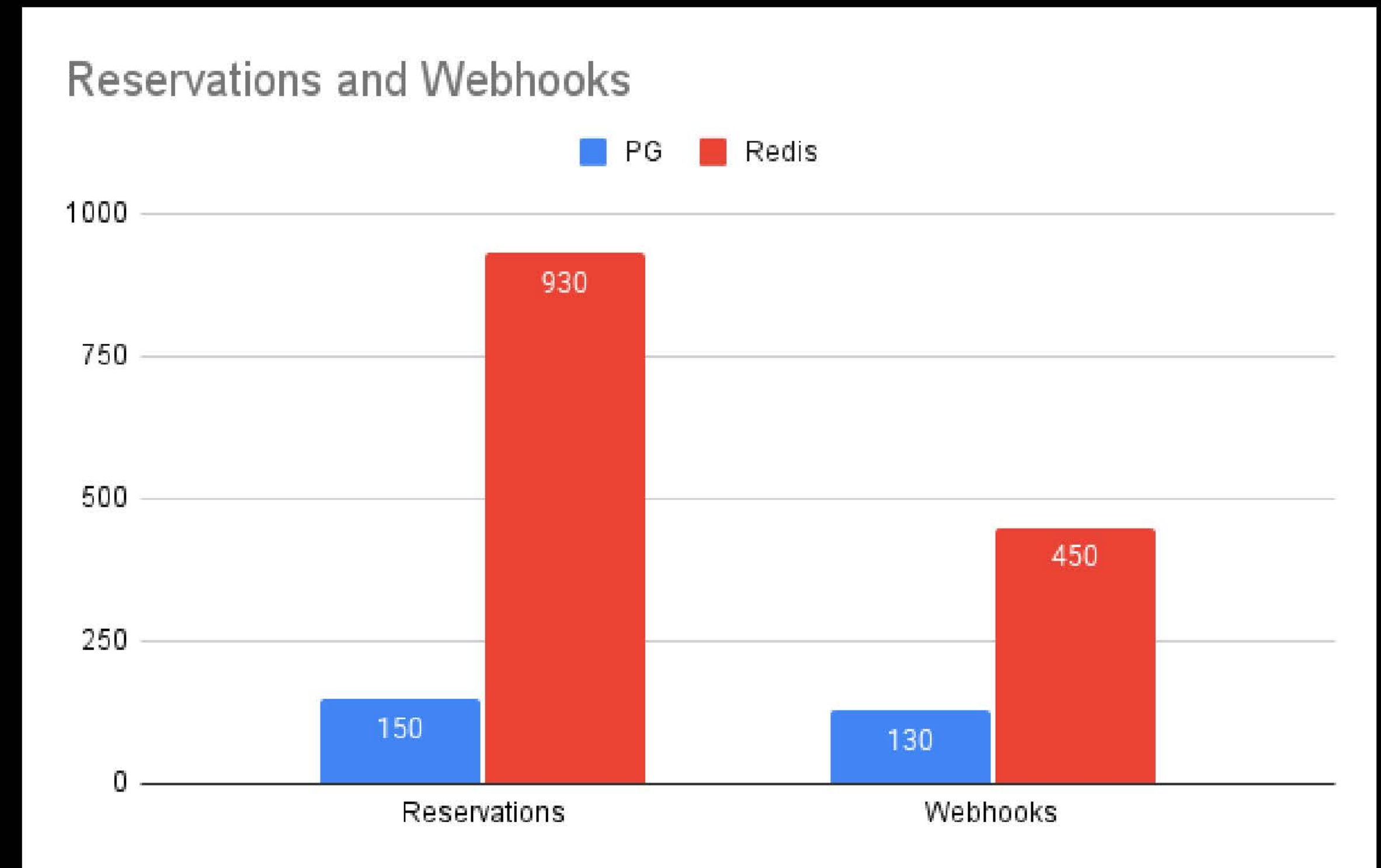
- Sessions
- Idempotency
- Orders

### Results:

- Reservations: **~930 ops/s**
- Webhooks: **~450 ops/s**

Great... except Redis **everysec** can **lose 1s of orders** on crash.

That's not acceptable.



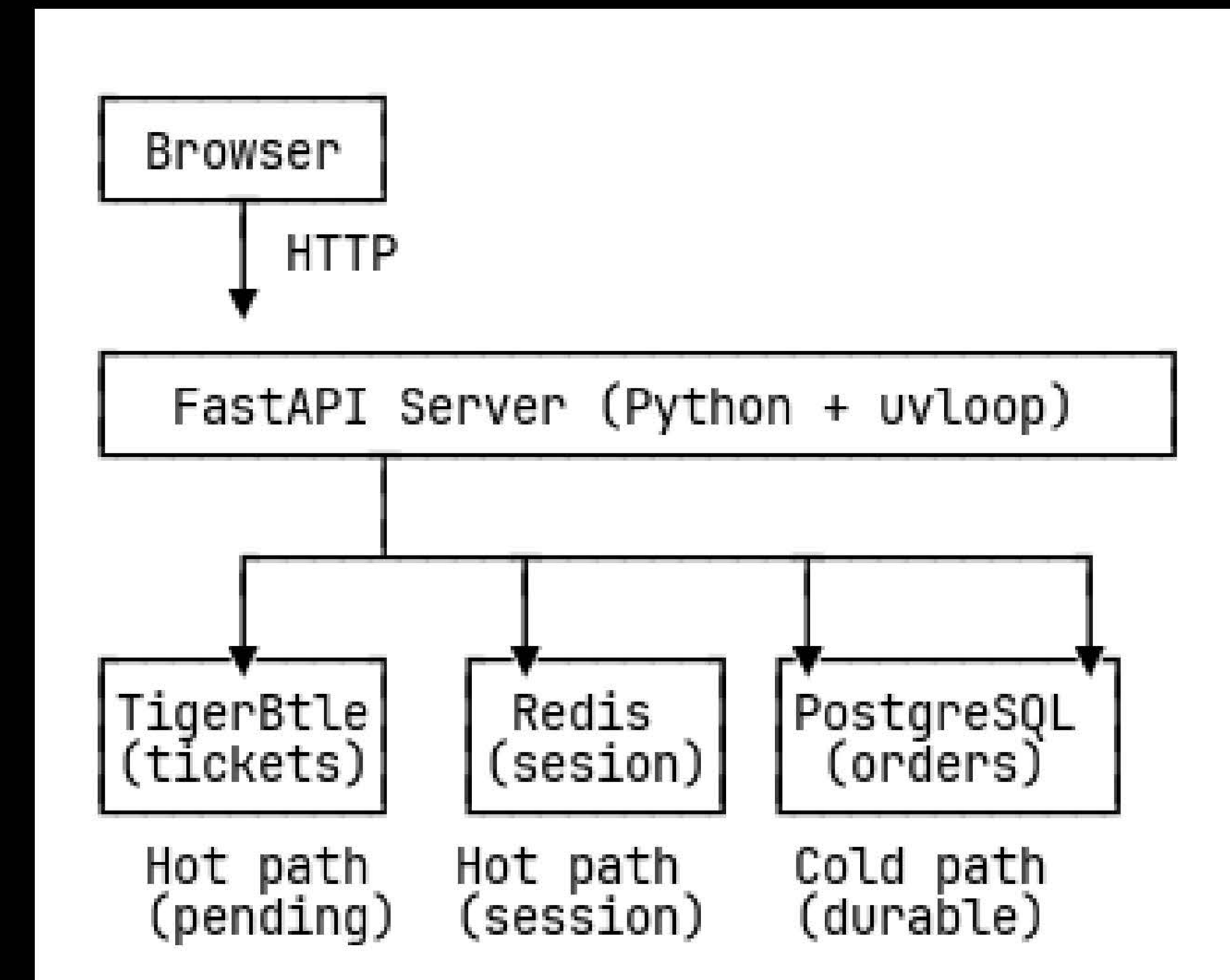


### Rafael's suggestion:

- Use **Redis** for **hot data**:
  - Payment sessions
  - Idempotency keys
- Use **PostgreSQL** only for **cold data**
  - Completed orders

### TigerBeetle:

- Always handles **ticket accounting**
- Always fully durable





# Three Configuration Levels



## Level 1 – PG Only

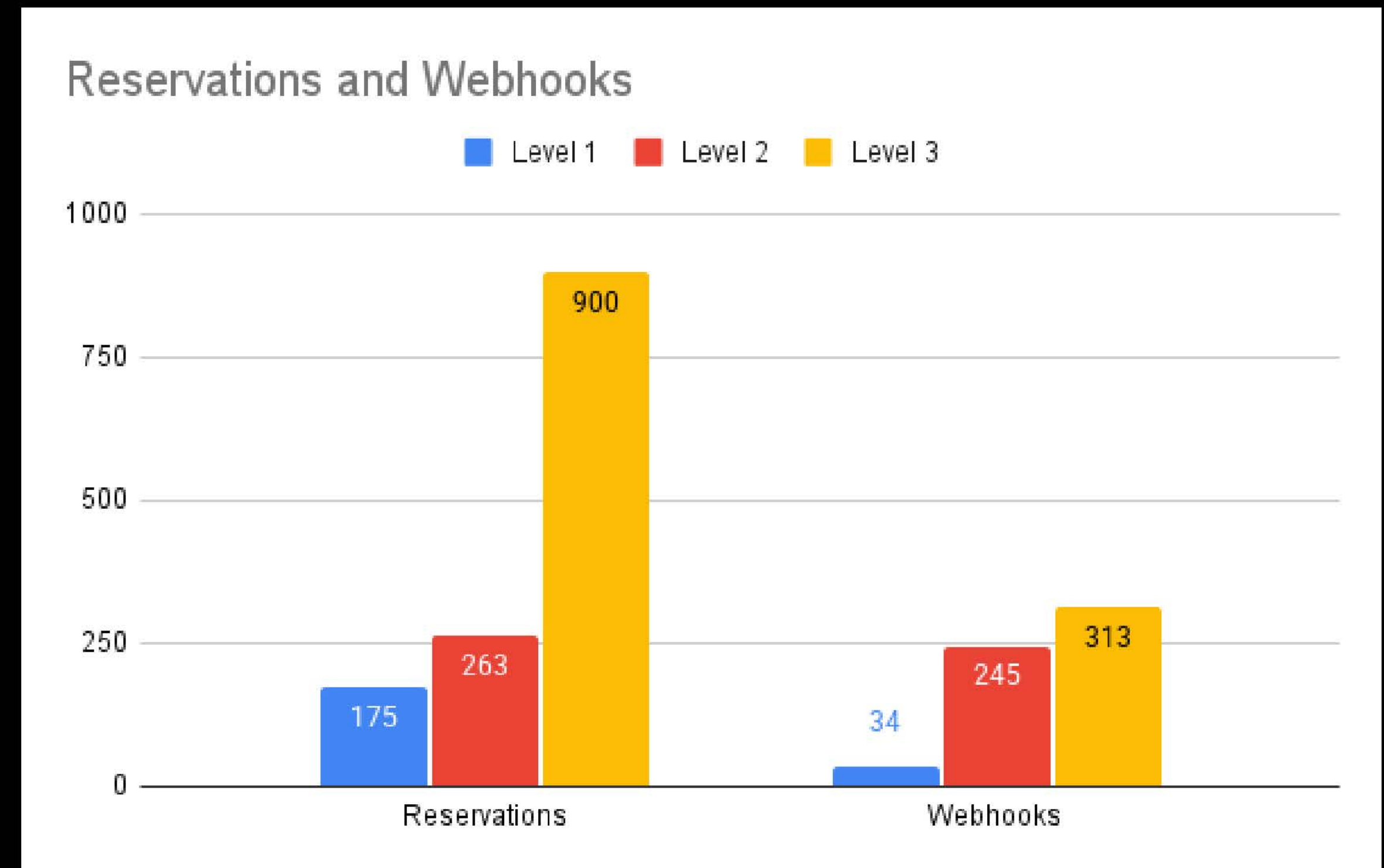
- PG for sessions, accounting, orders

## Level 2 – PG + Redis

- Redis: sessions
- PG: accounting + orders

## Level 3 – TB + Redis

- Redis: sessions
- TigerBeetle: accounting
- PG: orders





**TigerBeetle** loves **big batches**:

- Up to **8,190 transfers** per call
- One network round-trip amortized

**FastAPI** loves **per-request awaits**:

- Each request calls **create\_transfers()** once
- No chance to “wait a bit and batch”

Instrumentation showed:

- **Batch size  $\approx 1$**
- We were flying a 747 with 1 passenger per flight.



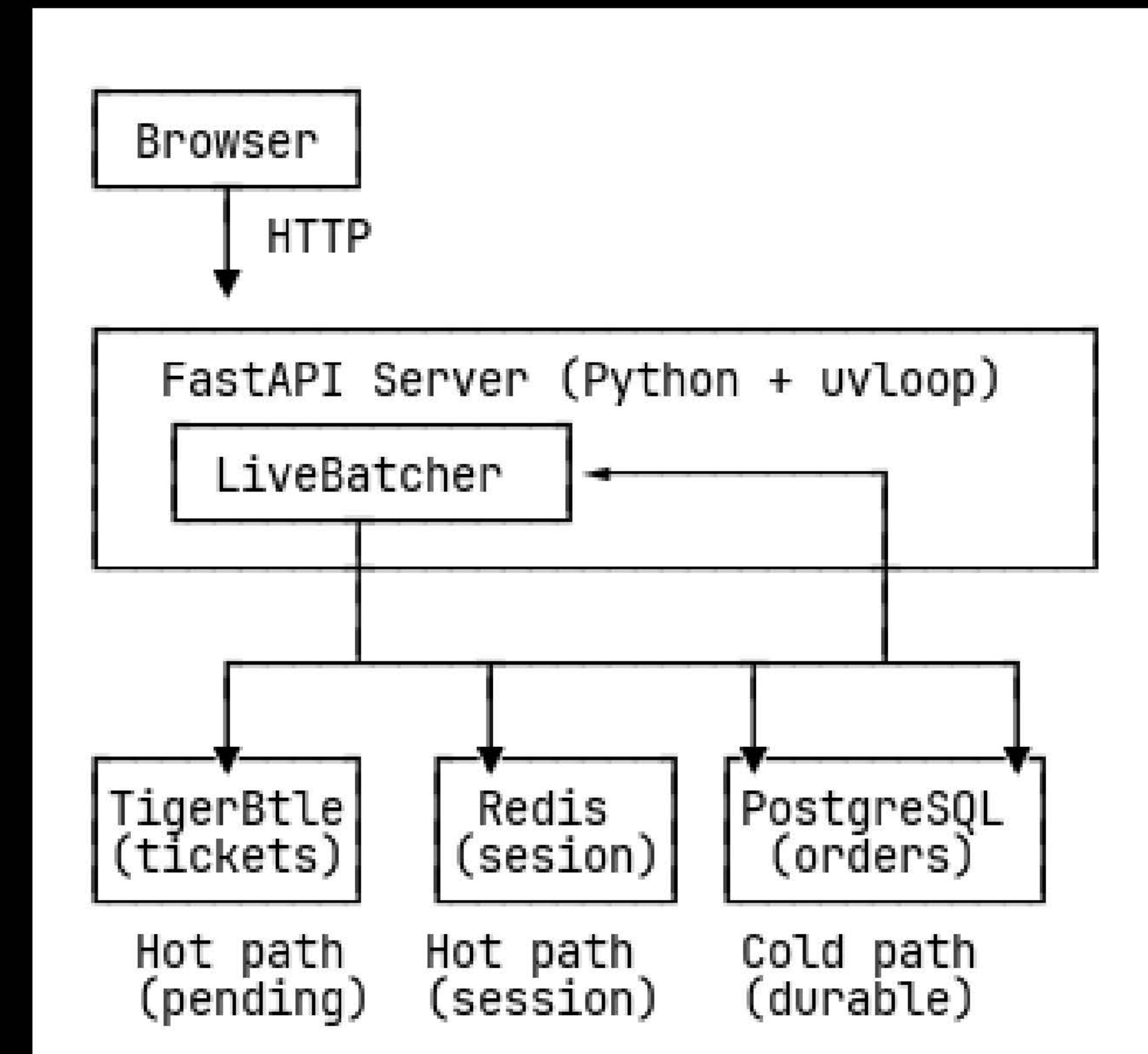


## LiveBatcher:

- Sits between app and TB client
- Collects concurrent requests
- Packs them into batches while previous batch is in flight
- Maps TB errors back to each caller

## Behavior:

- Under load: **batch sizes 5–6 transfers**
- Low load: small batches, low latency





With hot/cold path + LiveBatcher:

- Reservations: **977 ops/s**
- Average TB **batch size: 5–6**

Compared to baseline:

- PG-only: ~150 ops/s
- TB+Redis (before batching): ~900 ops/s

**LiveBatcher gave us the final 8%** — the earlier wins came from **architecture**, not micro-optimizations.



# The Single-Worker Paradox



## 1 worker

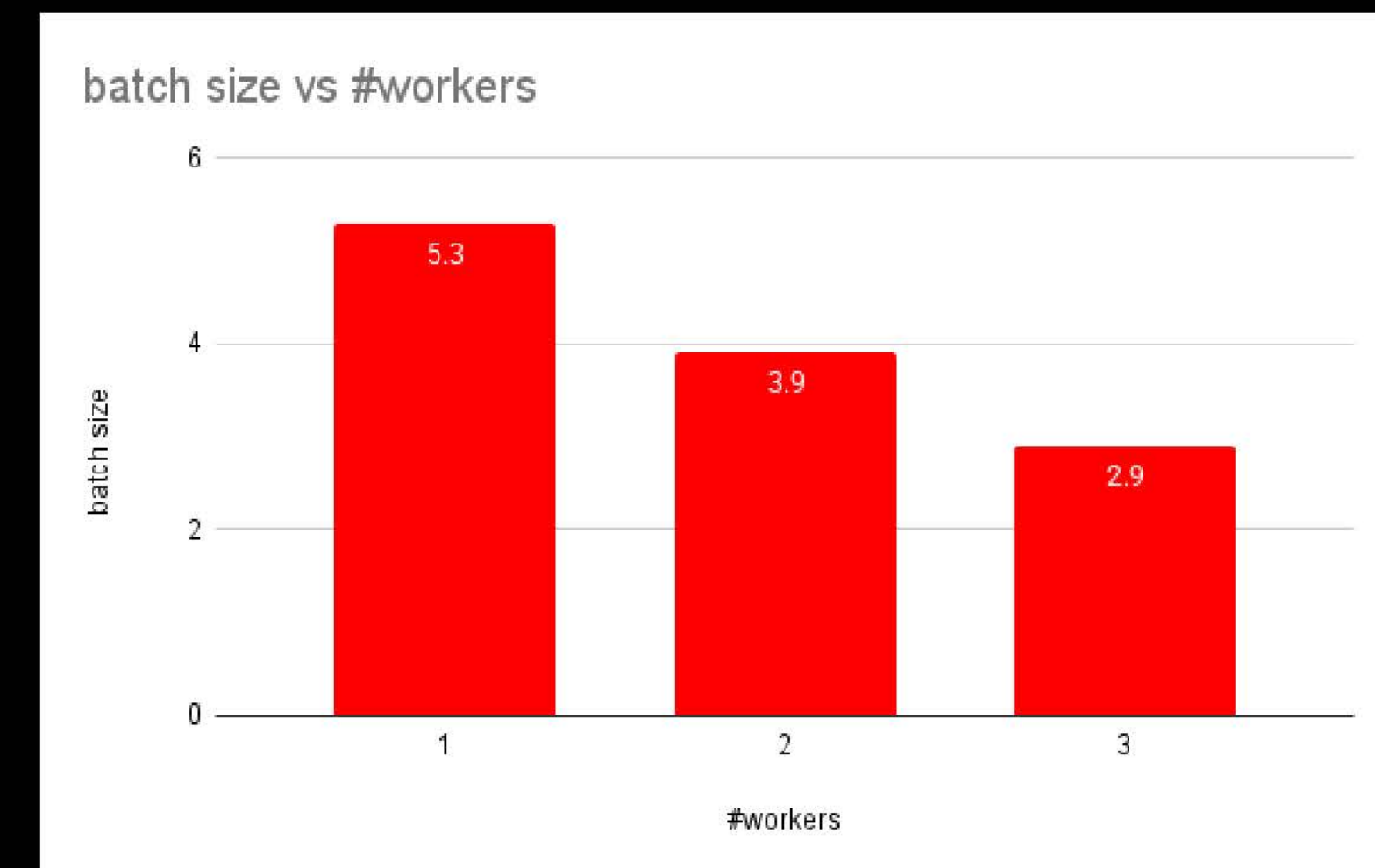
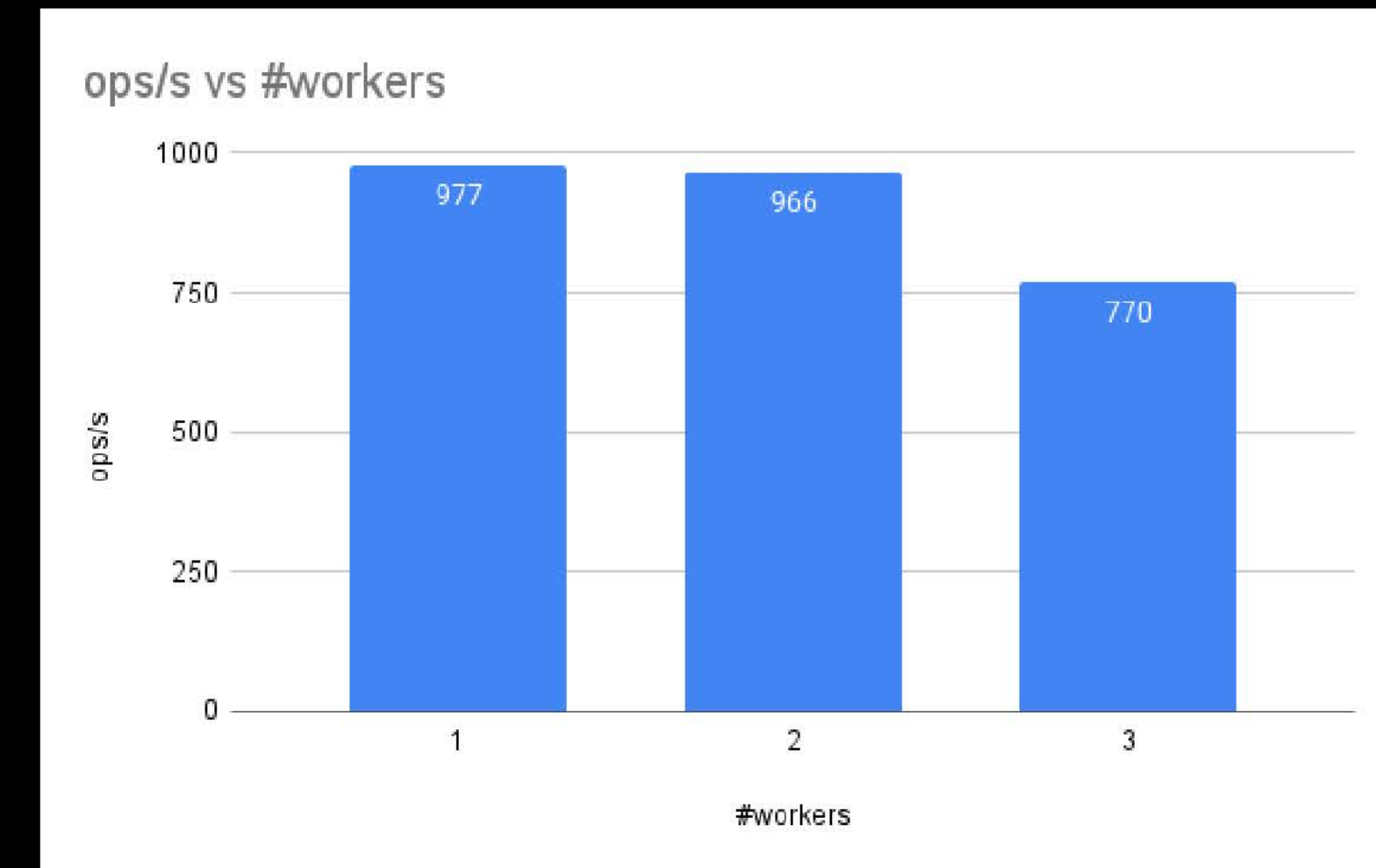
- 977 ops/s
- Avg batch size: 5.3

## 2 workers

- 966 ops/s
- Avg batch size: 3.9

## 3 workers

- 770 ops/s
- Avg batch size: 2.9



More workers -> **smaller batches** -> lower throughput.



**End result (Python, 1 worker):**

- **977 reservations / second**
- **$\approx 15\times$  Oasis baseline**
- **TB batch size: 5–6 transfers**

The real limit now isn't TigerBeetle  
— it's **Python's event loop overhead.**





### In checkout:

- We sped up **accounting** (PG TB) by  $\approx 3\times$
- Moved sessions to Redis
- But every request still spends:
  - ~5ms in Python/FastAPI
  - Routing, JSON, business logic

### Amdahl's Law says:

- That **serial 5ms** chunk becomes a **hard ceiling**
- Even with “infinite-speed” TB, Python caps us



If we keep the same **architecture**:

- Same double-entry model
- Same hot/cold split
- Same LiveBatcher idea

...but move to:

- Go, Zig, ...

Then:

- Python's 5ms -> maybe 0.1 – 0.5ms
- Batch sizes can grow
- Amdahl's Law says:
- 10–50× more throughput is realistic





## CHALLENGE:

- Same patterns:
  - Double-entry resource modeling
  - Hot/cold path
  - Auto-batching
  - Single-worker friendly design
- Any language, any stack
- Benchmark it, share results

<-- or do it your way :-)

<-- ... any platform? Raspberry Pi?

<-- please :-)

## Resources:

- [tigerfans.io](https://tigerfans.io) (demo + TigerBench)
- [github.com/renerocksai/tigerfans](https://github.com/renerocksai/tigerfans)



## **1. Ticket Booking**

- Start a checkout
- See **pending reservation** appear

## **2. MockPay**

- Complete payment
- Webhook posts TB transfers

## **3. Admin Live View**

- Orders, reservations, timeouts

## **4. TigerBench**

- Compare PG-only vs PG+Redis vs TB+Redis
- Ops vs. Transactions





# **TigerFans: High-Performance Ticketing with TigerBeetle**

**From Double-Entry Accounting to 977 Tickets per Second**

# **THANK YOU for your attention!**

## [Backup] Amdahl's Law: Component Breakdown



### Baseline checkout (Level 1 – PG Only):

- Total time  $\approx$  **15.4 ms**
- PG accounting: **14.03 ms** (~91%)
- PG sessions: **1.36 ms** (~9%)

### Optimized checkout (Level 3 – TB + Redis):

- Total time  $\approx$  **5.63 ms**
- TB accounting: **4.58 ms**
- Redis sessions: **1.04 ms**

### Amdahl's Law:

$$\text{Speedup\_overall} = 1 / ((1 - P) + P / S)$$

### Here:

- Accounting  $P \approx 0.91$ ,  $S \approx 3.0$
- Predicted speedup  $\approx$  **2.73×**
- Measured component speedup  $\approx$  **2.74×**



## [Backup] Amdahl & the Batching Ceiling



For batching, Amdahl applies again:

- TigerBeetle batch time  $\approx$  **3 ms**
- Python per-request serial work  $\approx$  **3 ms**
- Routing, parsing, validation
- Business logic
- Preparing TB transfers

During one TB batch ( $\sim 3$  ms):

- Python can push  $\approx 1$  new request to the batcher

So batch size naturally stabilizes around:

- **2–6 transfers** depending on concurrency load

In Go/Zig:

- Serial overhead  $\approx 0.1$  ms
- Same architecture  $\Rightarrow$  batches of **30–50+**