How I Learned to Stop Worrying and Love Raw Events

Event Sourcing & CQRS with FastAPI and Celery

PyCon Athens 2025

My Journey: From Celestial Chaos to System Chaos

- Senior Staff Engineer @ Orfium
- 10+ years Python experience
- Background: Physics → Computational Physics → Software Engineering
- Now: Debugging real-world system chaos
- Philosophy: Right architecture for each problem

The Nightmare: "Who Deleted My User?"

A real debugging story:

```
def delete_user(user_id: int):
   db.delete_user(user_id)
```

The problem:

Monday 3:47 PM: "Sarah's account is missing!"

Tuesday 9:15 AM: "When was it deleted? Who did it? Why?"

What we can't answer:

- X When was the user deleted?
- X Who deleted the user?
- X Why was it deleted?

The system has no memory of what happened

Enter Event Sourcing: The System That Remembers

We store every change as an immutable event:

```
UserDeleted(
    event_id=uuid4(),
    aggregate_id="user_123",
    revision=5,
    version=1,
    timestamp=datetime.now(),
    event_type="USER_DELETED",
    data={ "deleted_by": "admin_456", "reason": "Account closure request" }
)
```

Now we can answer everything:

- **When**: March 15, 3:47 PM
- Who: Admin ID 456
- Why: Account closure request

Every action becomes a permanent record

Core Concepts: Events

Immutable Facts

Events are immutable facts that represent state changes in the system.

Example:

User Created Event - John Doe, john@example.com, March 15

Key characteristics:

- Immutable: Once created, events never change
- Facts: They represent what actually happened
- Complete: Each event contains all necessary data
- Revisioned: Events have sequence numbers for ordering
- Versioned: Events have schema versions for serialization

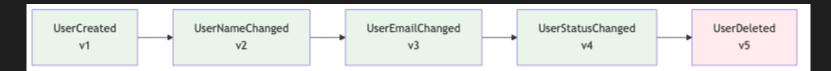
Key principle: Events are immutable facts - they never change

Core Concepts: Event Streams

Ordered Sequences

Event streams are ordered sequences of events for a specific aggregate.

Example:



Key characteristics:

- Ordered: Events have strict chronological ordering
- Complete: Contains the full history of an aggregate
- Replayable: Can rebuild any point in time
- Source of truth: The definitive record of what happened

The stream is the source of truth - rebuild any point in time

Core Concepts: Commands

Intent to Change

Commands represent the intent to change the system state.

Example:

"Create a new user account"

Key characteristics:

- Intent: They express what we want to happen
- Validation: Can be validated before execution
- Idempotent: Safe to retry if needed
- Entry point: The starting point for all changes

Commands are the entry point - they represent what we want to do

Core Concepts: Queries

Intent to Read (CQRS Separation)

Queries represent the intent to read data from the system.

Example:

"Show me user John Doe's profile"

Key characteristics:

- Read-only: They never change system state
- Optimized: Designed for specific read patterns
- Separate models: Different from command models (CQRS)

Queries are separate from commands - different models for different purposes

Core Concepts: Aggregates

Domain Logic

Aggregates contain domain logic and apply business rules to create events.

Example:

- User email must be unique
- Cannot delete already deleted user

Key characteristics:

- Business rules: Enforce domain-specific validation
- State management: Maintain current state from events
- Event creation: Generate new events based on commands

Aggregates apply business logic and create events

Core Concepts: Event Store

Source of Truth

Event Store is the append-only storage for all events in the system.

Example:

User John Doe's Event Stream

- Event 1: User Created (March 15, 2:30 PM)
- Event 2: Email Changed (March 16, 10:15 AM)

Key characteristics:

- Append-only: Events are never modified or deleted
- Immutable: Once written, events are permanent
- Stream management: Organizes events by aggregate

Event Store is append-only - events never change or delete

Core Concepts: Projections

Building Read Models

Projections build optimized read models from events for fast querying.

Example:

- Event: User Created → Action: Create user record
- Event: Email Changed → Action: Update email field

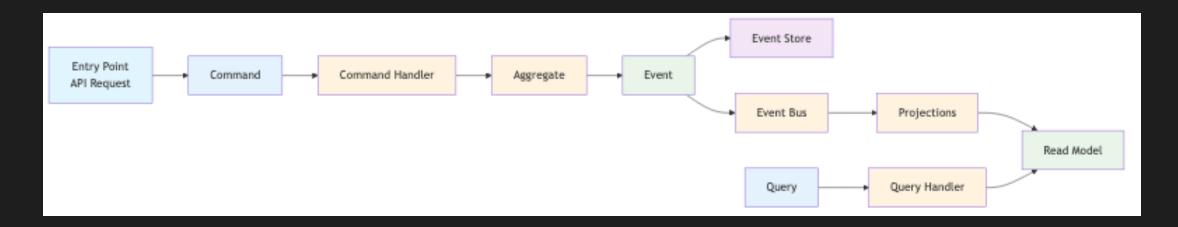
Key characteristics:

- Event-driven: Triggered by new events
- Read-optimized: Designed for specific query patterns
- Denormalized: Optimized for performance, not normalization
- Eventually consistent: Updated asynchronously

Projections update read models from events

How Everything Works Together

The complete flow:



Each interaction follows this pattern - from command to projection

FastAPI: The Command Interface

Real implementation with Pydantic:

```
@router.post("/users")
async def create_user(
    user_data: dict,
    handler: CreateUserCommandHandler = Depends(InfraFactory.create_user_command_handler)
):
    # Create command with validation
    command = CreateUserCommand(
        name=user_data["name"],
        email=user_data["email"]
)

# Process command
await handler.handle(command)

# Return immediately (event stored successfully) or catch exceptions via middleware
    return {"user_id": command.user_id}
```

FastAPI commands accept requests and return immediately after event storage

Command Handlers: Business Logic

How we structure command processing:

```
class CreateUserCommandHandler:
    async def handle(self, command: CreateUserCommand) -> None:
        # Retrieve all events for this aggregate
        events = await self.event store.get stream(command.user id)
        # Create empty aggregate and replay events
        user = UserAggregate(command.user id)
        for event in events:
            user.apply(event)
        # Call domain method and get new events
        new events = user.create user(command.name, command.email)
        # Persist and dispatch events using unit of work
        async with self.uow:
            await self.event_store.append_to_stream(command.user_id, new_events)
            await self.event handler.dispatch(new events)
```

Command Handler orchestrates: Event Store + Event Handler with Unit of Work

Event Handler: Celery Integration

How events are dispatched to Celery tasks:

```
class CeleryEventHandler:
    def init (self):
        # Map event types to Celery tasks
        self.event handlers = {
            "USER CREATED": [
                "process_user_created_task",
                "send welcome email task"
             ... other event types
    async def dispatch(self, events: List[Event]) -> None:
        for event in events:
            for task_name in self.event_handlers[event.event_type]:
                # All tasks receive the same event payload structure
                celery_app.send_task(task_name, kwargs={"event": event.model_dump()})
```

Event Handler dispatches to message queues, Celery tasks handle messages and call projections

Celery Tasks: Event Processing

How Celery tasks process events and call projections:

```
@app.task(name="process_user_created_task")
def process_user_created_task(event: Dict[str, Any]) -> None:
    """Celery task for processing USER_CREATED events"""
    # Get infrastructure factory
    factory = get_infrastructure_factory()

# Get projection
    projection = factory.create_user_created_projection()

# Process the event using async_to_sync
    async_to_sync(projection.handle)(EventDTO(**event))
    logger.info(f"Successfully processed USER_CREATED event for user {EventDTO(**event).aggregate_id}")
```

Celery tasks use async_to_sync to bridge async projections with sync Celery

Projections: Event-Driven Read Models

How projections build read models:

```
class UserCreatedProjection:
    async def handle(self, event: Event) -> None:
    # Build read model from event
    user_data = {
        "aggregate_id": event.aggregate_id,
        "name": event.data.get("name"),
        "email": event.data.get("email"),
        "status": event.data.get("status"),
        "created_at": event.timestamp,
    }

# Save to read model
    await self.db.save(user_data)
```

Projections update read models from events

FastAPI: Query Interface

How we expose read models:

```
@users_router.get("/{user_id}/")
async def get_user(
    user_id: str,
    query_handler: GetUserQueryHandler = Depends(InfraFactory.create_get_user_query_handler)
) -> Dict[str, Any]:
    return await query_handler.handle(GetUserQuery(user_id=user_id)).model_dump()

@users_router.get("/{user_id}/{timestamp}/")
async def get_user_at_timestamp(
    user_id: str,
    timestamp: datetime = Query(..., description="ISO 8601 format: YYYY-MM-DDTHH:MM:SSZ"),
    query_handler: GetUserAtTimestampQueryHandler = Depends(InfraFactory.create_get_user_at_timestamp_query_handler)
):
    return await query_handler.handle(GetUserAtTimestampQuery(user_id=user_id, timestamp=timestamp)).model_dump()
```

FastAPI queries expose read models

Eventual Consistency: The Real Challenge

The story: "Update user's first name"

```
# User updates first name
PUT /users/123/ {"first_name": "John"}
# API returns success immediately
# But read model might not be updated yet
```

Two approaches to handle this:

- 1. Optimistic Updates (Naive)
- 2. Outbox Pattern (Advanced)

Eventual consistency requires thoughtful UI design

Performance with Snapshots

The performance challenge:

```
async def handle(self, command: CreateUserCommand) -> None:
    events = await self.event_store.get_stream(command.user_id) # 10,000 events!
    user = UserAggregate(command.user_id)
    for event in events:
        user.apply(event) # Takes 5 seconds @
```

The solution: Snapshots in Command Handler

```
async def handle(self, command: CreateUserCommand) -> None:
    try:
        snapshot = await self.snapshot_store.get_latest_snapshot(command.user_id)
        recent_events = await self.event_store.get_events_since_snapshot(command.user_id, snapshot.revision)
        user = UserAggregate.from_snapshot(snapshot)
        for event in recent_events:
            user.apply(event)
    except SnapshotNotFound:
```

Snapshots require aggregate changes - rebuild state efficiently

Error Handling & Retries: Two Different Worlds

Commands (Synchronous) - API Failures:

```
# Unit of Work ensures atomicity
async with self.uow:
    await self.event_store.append_to_stream(user_id, new_events)
    await self.event_handler.dispatch(new_events)
# Either succeeds or fails - API gets 500
```

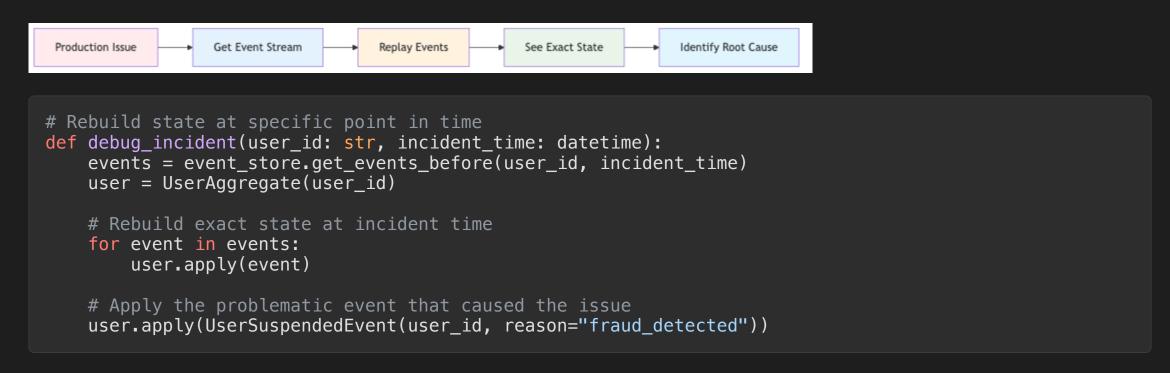
Projections (Asynchronous) - Celery Retries:

```
# Celery handles retries with late acknowledgment
@app.task(bind=True, max_retries=3, acks_late=True)
def process_user_created_task(self, event: Dict[str, Any]) -> None:
    projection.handle_user_created(event)
```

Different strategies for different failure modes

Debugging Superpowers: Testing Business Logic

The story: "What was the user's state at 3:47 PM?"



Test business logic with real production data

Summary: Key Takeaways

Start Simple

- No fancy event stores needed initially
- Familiar tools, powerful results

⚠ When NOT to use Event Sourcing

- Simple CRUD or high-frequency systems (immediate consistency needed)
- Teams without distributed systems experience

@ What you gain

- Complete audit trail & time travel
- Debugging superpowers with real production data
- Scalability with eventual consistency

Event sourcing + Python ecosystem = distributed systems superpowers

Thank You!

Q&A

Let's Connect!

GitHub: github.com/anmarkoulis

LinkedIn: linkedin.com/in/anmarkoulis

Dev.to: dev.to/markoulis