

ULDA Enterprise Architecture — Technical Reference

A Domain-Oriented Master Data & Customer Data Platform for Lead-360 on PostgreSQL

Data Architecture — InternetSP / ULDA v6.1

2026-07-16

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1. Executive Summary

The **Unified Lead Dataset Architecture (ULDA)** is an enterprise data platform that masters ~1.06 million deduplicated lead identities (from ~1.76 M raw source rows) into a single, governed, AI-ready system on a live production PostgreSQL database, with **zero downtime** during its construction.

Rather than a single wide “lead” table, ULDA is organised into **four cooperating domains**, each with a distinct change-velocity, storage strategy, and consumption pattern:

DOMAIN	NATURE	CHANGE VELOCITY	EXAMPLE OBJECTS
A — Master Data	Golden, mastered entities	Slow (SCD2)	identity_registry , person , company , address
B — Activity / Event	Append-only history	High	interaction , enrichment , order
C — Reference	Curated external facts	Medium	address_serviceability
D — Serving / CDP	Denormalised projections	Derived	mv_lead_360 , v_lead_journey_flat , lead_embedding

This separation is the core design decision. It is the difference between a spreadsheet and a data platform: identity is mastered once and never duplicated; activity is preserved forever as immutable history; and consumers (CRM, dialer, analytics, AI) read fast, purpose-built projections. This document specifies each domain and component in full, recommends the physical deployment, provides seven architecture diagrams, and closes with a critical assessment.

2. Architectural Overview

2.1 The four-domain model

ULDA applies a **medallion + Master Data Management (MDM) + Customer Data Platform (CDP)** pattern specialised for identity:

- **Master Data (A)** answers “*who is this?*” — one immutable golden identity per real person, with attribute-level survivorship and full temporal history.
- **Activity / Event (B)** answers “*what happened?*” — every interaction, enrichment, and order as an append-only, time-partitioned event.
- **Reference (C)** answers “*what do we know about the world?*” — serviceability/coverage facts keyed to addresses, sourced externally.
- **Serving / CDP (D)** answers “*give me the answer, fast*” — denormalised materialised views and vectors that consumers and AI read directly.

2.2 Why divide into four domains?

Because the four kinds of data have **fundamentally different physics**:

1. **Change velocity differs by orders of magnitude.** An identity's name changes rarely; its interactions arrive constantly. Storing both in one row forces every high-frequency event to lock and rewrite the slow golden record — creating contention, bloat, and lost history.
2. **Lifecycle and retention differ.** Golden identities are kept indefinitely and versioned (SCD2); events are partitioned by time and eventually archived; serving projections are disposable and rebuildable.
3. **Access patterns differ.** Master data is point-lookup by key; events are range scans by time; serving is wide denormalised reads. Each wants a different indexing/partitioning strategy — impossible to optimise in one table.

4. **Governance differs.** PII concentrates in Master Data; consent/DNC lives in Events; serving must be *masked* for AI. Domain boundaries make governance enforceable.

2.3 Why this beats “one big table”

A single wide lead table (the intuitive design) fails on every enterprise axis:

CONCERN	ONE BIG TABLE	FOUR-DOMAIN MODEL
Journey history	capped at fixed columns; overwrites	unlimited, immutable events
Normalisation	update/insert/delete anomalies	3NF master + append-only events
NULL density	most columns null for most rows	data exists only where relevant
Indexing	one index strategy for all needs	per-domain optimal indexes
Partitioning	can't partition identity by event time	events partitioned by time
Concurrency	every event UPDATES the golden row	INSERT-only events, no contention
AI features	mixes stable + volatile → drift/leakage	clean per-domain feature store
Scale to 1B	row & index bloat	event volume scales independently

2.4 MDM alignment

ULDA implements the canonical MDM disciplines: a **single source of truth**

(`identity_registry`), **survivorship** (golden record chooses the best value per attribute with lineage), **crosswalk** (`source_crosswalk` maps one golden ID to many source system IDs), **match/merge with audit** (deterministic + probabilistic resolution, reversible merges), and **stewardship** (a governance schema holding data-quality findings and a human match-review queue).

2.5 CDP alignment

A CDP unifies customer data from many sources into a persistent, single customer view and activates it to operational tools. ULDA's Serving domain is exactly this: `mv_lead_360` is the unified profile; `v_lead_journey_flat` is the cross-channel timeline; the Reverse-ETL surface activates segments back to Warroom / SMS / dialer under a consent gate.

3. Physical Database Design

3.1 The question

Should ULDA be **one database with many schemas**, **many databases**, **many servers**, or a **hybrid**?

3.2 Recommendation: one PostgreSQL database, multiple schemas (with a replica for scale-out reads)

```
PostgreSQL instance (database: leads_db)
|
├─ ulda_master      – Master Data domain (A)
├─ ulda_events      – Activity / Event domain (B)    [partitioned]
├─ ulda_ref         – Reference domain (C)
├─ ulda_serve       – Serving / CDP domain (D)        [materialised views +
pgvector]
└─ ulda_gov         – Governance (DQ, consent, lineage, match-review, audit)
    (public.*       – legacy operational tables, authoritative during
transition)
```

This is the deployment ULDA runs today, and it is the correct choice at this scale.

3.3 Why schemas, not separate databases

- **Cross-domain queries need joins.** `mv_lead_360` joins master + events + reference in a single query. Across separate PostgreSQL databases those joins are impossible without foreign-data wrappers (slow, fragile). Schemas give you one transaction boundary, one query planner, and real foreign keys across domains.

- **Referential integrity.** FKs from every domain to `ulda_master.identity_registry` are only enforceable inside one database. Separate databases would push integrity into application code — a step backwards.
- **Operational simplicity.** One backup, one PITR timeline, one connection pool, one set of credentials/roles. Separate databases multiply operational surface for no benefit at $\leq$ low-billions of rows.
- **Security is still strong.** Schema-level `GRANT`s + distinct roles (`ulda_web` read-only, `ulda_writer`, `ulda_admin`) provide isolation without physical separation.

3.4 Why not multiple servers (yet)

A single well-provisioned node (here: 4 vCPU / 62 GB) comfortably serves ~1–2 M identities and their events. **Sharding across servers adds distributed-transaction complexity that is not justified until a single node is genuinely saturated.** The correct first scale-out step is a **read replica** (streaming replication) that offloads analytics/AI/serving reads from the OLTP primary — a hybrid that keeps one logical database while separating read and write load.

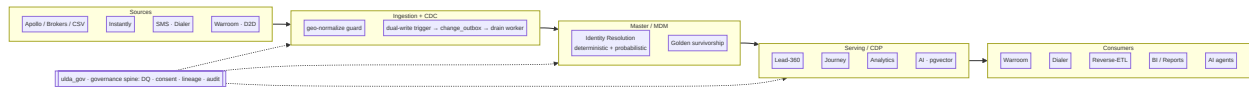
3.5 The recommended production topology (hybrid, staged)

1. **Now:** single primary, five schemas, continuous WAL archiving + PITR, verified off-host backups.
2. **Next (scale reads):** add a streaming **read replica**; route analytics, `lead_embedding` builds, and heavy `mv_*` refreshes there.
3. **Later (scale writes, > ~1 B events):** partition-level sharding of the Event domain (by time, then hash) and/or a dedicated OLAP tier (columnar) fed by CDC — without changing the logical four-domain model.

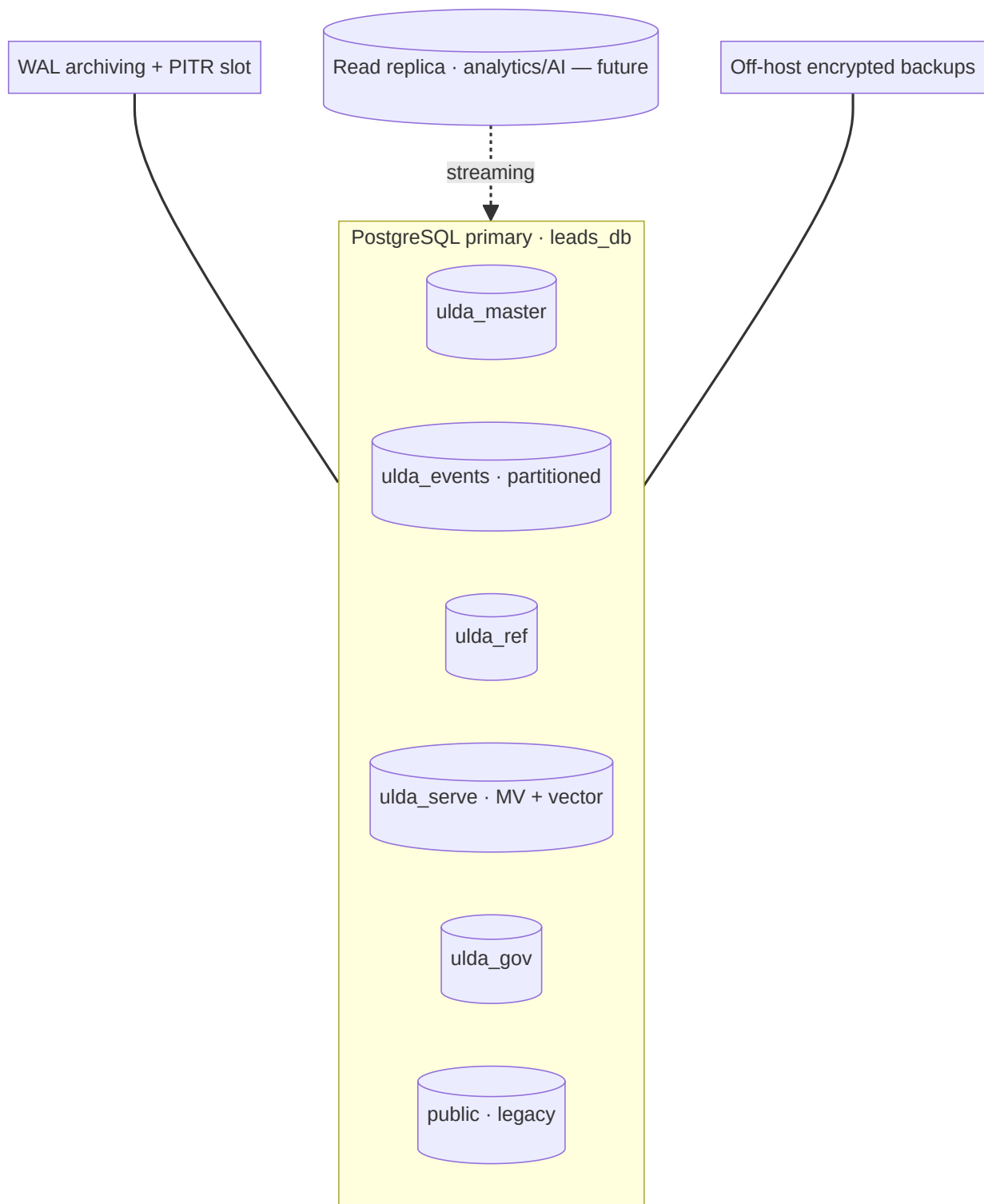
Principle: keep it one logical database until the workload — not the diagram — forces a split. Physical topology should follow measured load, not aesthetics.

4. Diagrams

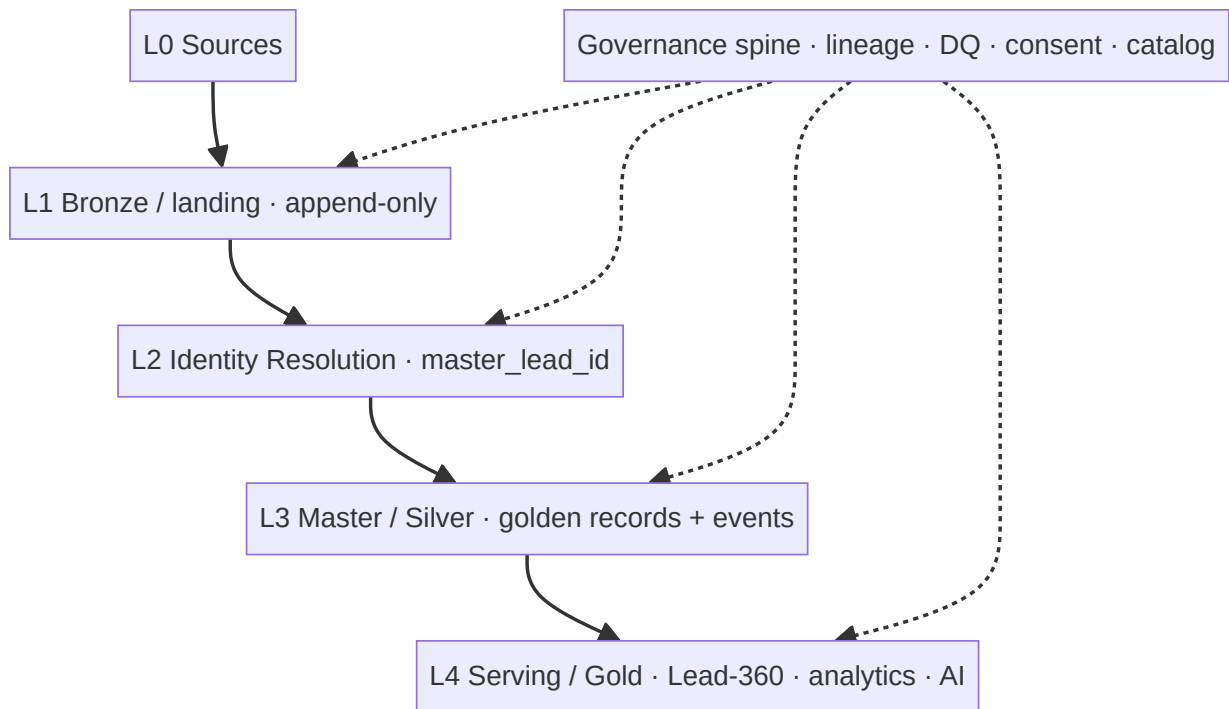
4.1 Enterprise System Architecture



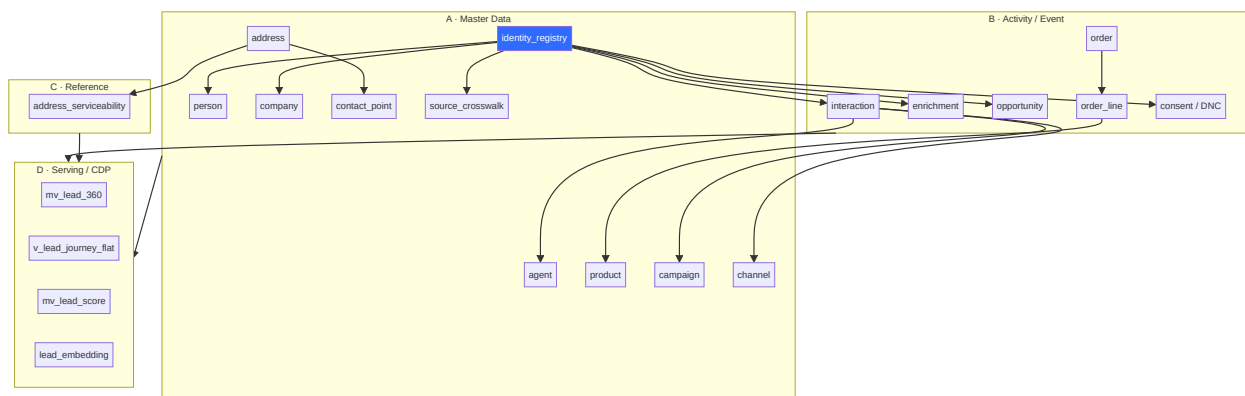
4.2 Physical Database Architecture



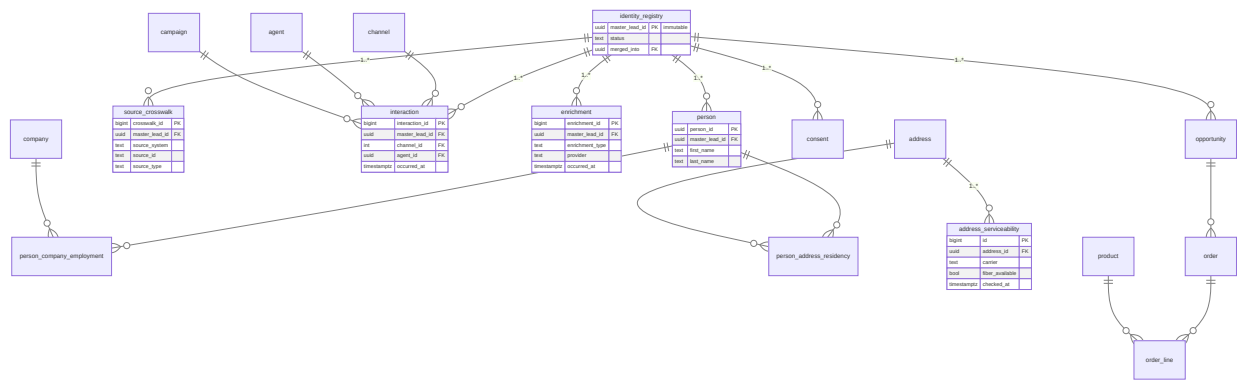
4.3 Logical Architecture



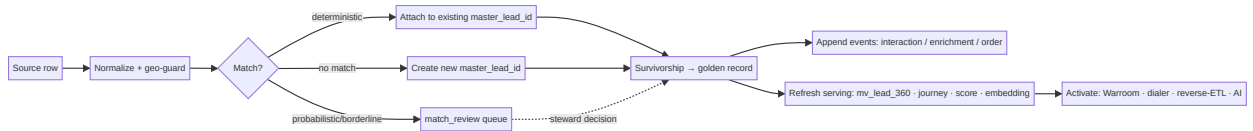
4.4 Master Data Domain Diagram



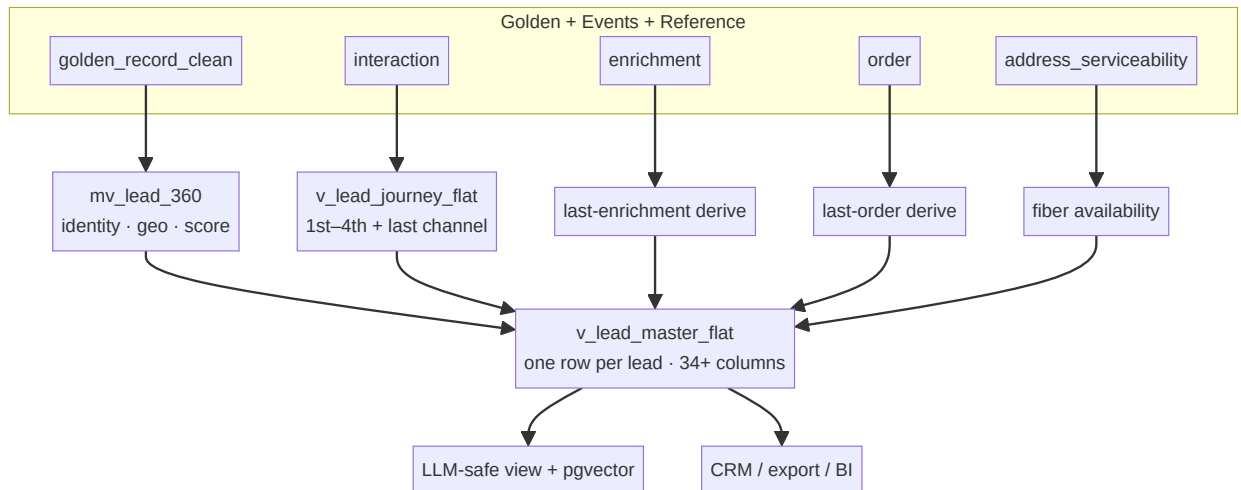
4.5 Entity Relationship Diagram (ERD)



4.6 Data Flow Diagram



4.7 Lead-360 Architecture



5. Domain A — Master Data

Components: `identity_registry`, `person`, `company`, `address`, `contact_point`, `source_crosswalk`, `agent`, `product`, `campaign`, `channel`.

1. **Purpose.** Hold one authoritative, deduplicated golden record per real-world entity (person, company, address, etc.), each with a permanent surrogate key.
2. **Business Objective.** Guarantee *One Lead = One Identity* so every downstream metric, message, and sale attributes to a single, correct customer — eliminating double-contact, wasted spend, and compliance risk.
3. **Responsibilities.** Assign/maintain immutable IDs; arbitrate conflicting attribute values (survivorship); maintain temporal history (SCD2); map source keys to golden keys (crosswalk); enforce referential integrity for all other domains.
4. **Why this domain exists.** Sources disagree and duplicate. Without a mastered layer, “the customer” is smeared across many rows keyed differently in each system. This domain is the single point where identity is resolved and truth is decided.
5. **Design Principles.** Surrogate immutable keys; append-then-supersede (never destructive update); attribute-level lineage; deterministic-first matching; separation of identity from activity.
6. **Main Components.** Identity spine (`identity_registry`), core entities (`person`, `company`, `address`, `contact_point`), dimensions (`agent`, `product`, `campaign`, `channel`), and the `source_crosswalk` bridge.
7. **Data stored.** Names, firmographics, normalized addresses, typed contact points, the golden survivorship values, and the source-to-master mapping — *not* events.
8. **Relationships.** It is the hub: every Event and Reference row references `identity_registry.master_lead_id` (or `address_id`); Serving projections read from it.
9. **Typical tables.** `identity_registry`, `person`, `company`, `address`, `contact_point`, `source_crosswalk`, `agent`, `product`, `campaign`, `channel` (all **master/dimension tables**).
10. **Example record.** `identity_registry(master_lead_id='7f3a...', status='active', primary_person_id='a91l...')` → `person(first_name='Reginaldo', last_name='G.')` → `source_crosswalk(source_system='instantly', source_id='998172', source_type='cold_email')`.
11. **Typical workflows.** New source row → normalize → match → create/attach master ID → recompute survivorship. Merge of two masters → `merged_into` set, audited, reversible.
12. **Enterprise best practices.** Golden-record survivorship with provenance; UNIQUE (`source_system`, `source_id`); deterministic before probabilistic; human review for low-confidence merges; SCD2 with `valid_from/valid_to/is_current`.
13. **Benefits.** True dedup; stable joins; complete lineage; backward compatibility (legacy keys preserved in crosswalk).

14. **Challenges.** Match precision vs recall; survivorship rule maintenance; merge/unmerge correctness; steward workload.
15. **Scalability.** Bounded at identity count (~1–2 M); B-tree point lookups; trigram indexes for fuzzy match; no partitioning needed.
16. **Security.** Highest PII concentration → strictest `GRANT` s; column-level PII classification; masking for non-privileged roles.
17. **Data governance.** Lineage per attribute; match-review queue; DQ findings flagged, not silently altered; DSAR-erasable flags.
18. **Performance.** Blocking keys to bound match comparisons; partial/covering indexes; survivorship materialised into `golden_record_full`.
19. **Future extensibility.** Household/company graphs, external identity graph enrichment, ML-based matching swapped in behind the same registry.

Components (Domain A)

- **identity_registry** — *Master table*. The identity spine: one row = one permanent golden `master_lead_id` (uuid, `gen_random_uuid()`). Exists so identity is generated once and never reused; merges recorded via `merged_into`. Every other domain FKs to it. Apps treat it as *the* lead ID; AI uses it as the join key for features. Schema: `(master_lead_id uuid PK, status text, merged_into uuid FK, primary_person_id uuid, canonical_hash text, created_at, updated_at)`.
- **person** — *Master table (SCD2)*. Human attributes tied to a master ID. Stores names; versioned over time. Apps display it; AI derives name/role features. `(person_id uuid PK, master_lead_id uuid FK, first_name, last_name, full_name, valid_from, valid_to, is_current)`.
- **company** — *Master table (SCD2)*. Firmographics (legal name, domain, industry, size/revenue band). Linked to persons via employment. `(master_company_id uuid PK, legal_name, domain, industry, size_band, revenue_band, is_current)`.
- **address** — *Master table (SCD2)*. Normalized postal locations with `normalized_hash` for dedup; the anchor for serviceability. `(address_id uuid PK, line1, line2, city, state, postal_code, country, normalized_hash)`.
- **contact_point** — *Master table*. Typed reachable endpoints (phone/email) as first-class children, each with source and validity — this is what lets phone-keyed and email-keyed sources collapse into one person. `(contact_point_id, master_lead_id FK, kind, value, status, source)`.
- **source_crosswalk** — *Master/bridge table*. Maps one golden ID to many `(source_system, source_id, source_type)` rows — the MDM heart and dedup ledger. `(crosswalk_id bigint PK, master_lead_id FK, source_system, source_id, source_type, match_method, confidence)`.

- **agent** — *Dimension/master table*. Sales/rep directory (name, role, team, territory, external logins). Referenced by interactions and orders. (agent_id uuid PK, full_name, email, role, team_id, territory_id, external_ids jsonb, status) .
- **product** — *Dimension table*. Sellable fiber plans/offers (carrier, speeds, price, term, offer_code). “Offer” is a product packaging, not a separate table. (product_id PK, carrier, name, download_mbps, upload_mbps, monthly_price, term_months, offer_code) .
- **campaign** — *Dimension table*. Marketing/outreach campaigns per channel. Referenced by interactions and memberships. (campaign_id text PK, name, channel_id FK, kind, status, start_at, end_at) .
- **channel** — *Reference/dimension table*. Controlled vocabulary of interaction channels (SMS, Email, Voice, Door, Web, Mail); one definition reused by interaction + campaign. (channel_id int PK, name, kind, provider) .

6. Domain B — Activity / Event

Components: interaction, enrichment, opportunity, order, order_line, consent / DNC .

1. **Purpose.** Record everything that *happens* to a lead as immutable, time-ordered events.
2. **Business Objective.** Enable attribution, journey analytics, compliance, and revenue tracking from a complete, trustworthy history.
3. **Responsibilities.** Append events (never overwrite); preserve chronology; link each event to its master identity, channel, agent, campaign, product; support retention/partitioning.
4. **Why this domain exists.** Events are high-velocity and unbounded; keeping them out of the golden row prevents lock contention, preserves history, and makes time-series analytics and partitioning possible.
5. **Design Principles.** Append-only; immutable; partition by event time; surrogate + time composite PK; foreign keys to master dimensions.
6. **Main Components.** interaction (touches), enrichment (data-provider events), opportunity / order / order_line (deal & revenue), consent /DNC (compliance state changes).
7. **Data stored.** Timestamped touches with channel/agent/campaign/disposition; enrichment provenance; deal stages and line items; consent grants/revocations.

8. **Relationships.** Every event FKs to `identity_registry` ; interactions FK to `channel` / `agent` / `campaign` ; order lines FK to `product` . Serving derives “first/last/journey” from these.
9. **Typical tables.** `interaction` (**transaction, partitioned**), `enrichment` (**transaction, partitioned**), `opportunity` , `order` , `order_line` , `consent` (all **transaction tables**).
10. **Example record.** `interaction(master_lead_id='7f3a...', channel_id=1 /*SMS*/, agent_id='...', campaign_id='C2UUBR7', interaction_type='sms_out', occurred_at='2026-07-08T14:03Z', software_type='sms_platform')` .
11. **Typical workflows.** Outbound SMS sent → interaction row. Provider returns phone → enrichment row. Rep books install → opportunity → order → order_line. Recipient texts STOP → consent revocation event → serving suppresses.
12. **Enterprise best practices.** Event sourcing; monthly RANGE partitioning; `pg_partman` auto-rollout; BRIN on time; never mutate history — correct with compensating events.
13. **Benefits.** Full auditability; accurate multi-touch attribution; compliance evidence; no contention on the golden record.
14. **Challenges.** Volume growth; partition management; event schema evolution; deduping duplicate webhooks.
15. **Scalability.** Volume scales independently of identity count; partition pruning + retention keep hot data small; shard by time then hash at extreme scale.
16. **Security.** Message bodies/PII in `metadata` → restricted; consent is a compliance-critical, tamper-evident record.
17. **Data governance.** Immutable audit trail; consent lineage; retention policy per event type; TCPA/DNC enforcement at read time.
18. **Performance.** Partition pruning; `(master_lead_id, occurred_at DESC)` for per-lead reads; covering indexes for hot analytic slices.
19. **Future extensibility.** New channels/event types add rows, not columns; stream to a real-time bus; feed ML training sets directly.

Components (Domain B)

- **interaction** — *Transaction table (partition by `occurred_at`)*. Every cross-channel touch. Exists to hold the unbounded journey that the flat design could not. Relates to identity + channel + agent + campaign. Apps read the timeline; AI derives recency/frequency/sequence features; the journey view pivots the first N + last touches from it. Adds `agent_login` + `software_type` (dialer/warroom/sms_platform) for the requested agent-activity trail. Schema: `(interaction_id bigint, master_lead_id uuid FK, channel_id int FK, agent_id uuid FK, campaign_id text FK, direction, interaction_type, disposition, agent_login, software_type, occurred_at`

```
timestampz, metadata jsonb, PRIMARY KEY(interaction_id, occurred_at)) PARTITION  
BY RANGE(occurred_at) .
```

- **enrichment** — *Transaction table (partitioned)*. One row per provider enrichment (Apollo/LeadMagic/IPQS/SARA). Exists so *full* enrichment history is preserved (not just “last”); “last enrichment date/type/source” is a cheap `max(occurred_at)` derivation. `(enrichment_id, master_lead_id FK, enrichment_type, provider, source_type, status, occurred_at, detail jsonb)` .
- **opportunity** — *Transaction table*. A qualified deal in the pipeline linking lead → product → owning agent → campaign. Feeds forecasting and conversion analytics. `(opportunity_id uuid PK, lead_id FK, person_id FK, account_id FK, product_id FK, owner_agent_id FK, campaign_id FK, stage, amount)` .
- **order** — *Transaction table*. A won sale (install) with amount and install date. `(order_id uuid PK, account_id FK, opportunity_id FK, agent_id FK, status, total_amount, ordered_at, install_date)` .
- **order_line** — *Transaction table*. Line items of an order, each referencing a `product` . Normalises multi-product orders. `(order_line_id PK, order_id FK, product_id FK, qty, unit_price)` .
- **consent / DNC** — *Transaction/compliance table*. Consent grants and revocations per channel — the fail-closed gate for SMS/dialer (TCPA). `(consent_id, master_lead_id FK, person_id FK, channel_id FK, status, basis, effective_at)` .

7. Domain C — Reference

Component: `address_serviceability` (AT&T, Kinetic, Spectrum, ...).

1. **Purpose.** Store curated external facts about the world — here, which carrier/technology is serviceable at an address, over time.
2. **Business Objective.** Sell the right product to the right address; a fiber ISP lives or dies by accurate serviceability.
3. **Responsibilities.** Maintain per-address, per-carrier availability with technology, speed, source, and check timestamp; supersede stale checks.
4. **Why this domain exists.** Serviceability is *address-scoped and multi-valued* (many carriers, changing over time) — it is neither identity nor activity, so it deserves its own reference domain rather than being flattened into a single “fiber” column.

5. **Design Principles.** Keyed to `address_id` ; temporal (`checked_at`); multi-carrier rows; source-attributed; curated (validated before trust).
6. **Main Components.** `address_serviceability` , reconciling the existing `ulda_ref.v_address_serviceability` view + the SARA/Kinetic/FCC checkers.
7. **Data stored.** (`address`, `carrier`, `fiber_available`, `technology`, `max_download_mbps`, `checked_at`, `source`) .
8. **Relationships.** FK to `ulda_master.address` ; consumed by Serving to surface “Fiber Availability” on Lead-360 and to prioritise dialer/canvass lists.
9. **Typical tables.** `address_serviceability` (**reference table**); `v_address_serviceability` (**standard view**).
10. **Example record.** (`address_id='...'`, `carrier='att'`, `fiber_available=true`, `technology='fiber'`, `max_download_mbps=5000`, `source='sara'`, `checked_at='2026-07-10'`) .
11. **Typical workflows.** Address checker runs → upsert serviceability row → Lead-360 shows fiber=yes → lead routed to the fiber campaign/dialer.
12. **Enterprise best practices.** Idempotent upserts; keep history of checks; source + confidence; separate reference data from operational identity.
13. **Benefits.** Accurate targeting; fewer wasted contacts; reusable across leads at the same address.
14. **Challenges.** Coverage-data freshness; carrier API variance; address-matching to the reference.
15. **Scalability.** Bounded by unique addresses; index by (`address_id`) and (`carrier`, `fiber_available`) .
16. **Security.** Low PII, but competitive-sensitive; standard read grants.
17. **Data governance.** Source lineage + `checked_at` freshness; supersede rather than overwrite.
18. **Performance.** Small, cache-friendly; join to Lead-360 via `address_id` .
19. **Future extensibility.** Add competitors, pricing, promo windows, install-feasibility (MDU/SFU), FCC/BDC datasets.

Component (Domain C)

- **address_serviceability** — *Reference table*. Per-address × carrier × time serviceability. Exists to model “Fiber Availability (AT&T, Kinetic, Spectrum...)” correctly as many rows, not one column. Relates to `address` ; consumed by Lead-360 and lead routing. Apps show availability; AI can feature-encode “fiber_available_att” etc. (`serviceability_id` bigint PK, `address_id` uuid FK, `carrier` text, `fiber_available` bool, `technology` text, `max_download_mbps` int, `checked_at` timestamptz, `source` text, `UNIQUE(address_id, carrier, checked_at)`) .

8. Domain D — Serving / Customer Data Platform (CDP)

Components: `mv_lead_360`, `v_lead_journey_flat`, `mv_lead_score`, `lead_embedding` (pgvector).

1. **Purpose.** Present fast, denormalised, purpose-built projections of the mastered data for applications, analytics, and AI.
2. **Business Objective.** Give every consumer a single, instant, correct customer view (Lead-360) without paying the cost of joining the normalised model on every read.
3. **Responsibilities.** Materialise the unified profile; compute the cross-channel journey; score/tier leads; expose LLM-safe features and vectors; activate segments (reverse-ETL) under consent.
4. **Why this domain exists.** OLTP-normalised models are optimal for writes and integrity but slow for wide reads; the Serving domain trades storage for read speed and shields the master from analytical load.
5. **Design Principles.** Derived (never a source of truth); rebuildable; refreshed `CONCURRENTLY` (no read blocking); PII-masked for AI; consent-gated for activation.
6. **Main Components.** `mv_lead_360` (profile), `v_lead_journey_flat` (timeline pivot), `mv_lead_score` (scoring), `lead_embedding` (semantic vectors), analytics MVs, activation views.
7. **Data stored.** Flattened identity+geo+score per lead; pivoted first-4+last channels; score/tier; 384-dim embeddings; rollups.
8. **Relationships.** Reads from Master + Events + Reference; writes to nothing authoritative; feeds Warroom, dialer, BI, AI, reverse-ETL.
9. **Typical tables.** `mv_lead_360`, `mv_lead_score`, `mv_analytics_*`, `lead_embedding` (**materialised views / derived tables**); `v_lead_journey_flat`, `v_lead_master_flat`, `v_lead_llm_safe` (**standard views**).
10. **Example record.** `mv_lead_360(master_lead_id='7f3a...', name='Reginaldo G.', street='7602 riptide dr', state='TX', zip='77072', completeness=0.7, source_systems={instantly}, source_count=1)`.
11. **Typical workflows.** Drain reconciles → `REFRESH MATERIALIZED VIEW CONCURRENTLY mv_lead_360` → apps/AI read the fresh profile; nightly analytics refresh; batch embedding on the replica.
12. **Enterprise best practices.** Concurrent refresh; unique index for concurrency; separate OLAP read path; semantic layer (governed metric definitions); feature store for ML.

13. **Benefits.** Sub-second wide reads; consistent single view; AI-ready; analytics isolated from OLTP.
14. **Challenges.** Refresh scheduling/freshness lag; MV storage; embedding compute cost; keeping derivations in sync with source.
15. **Scalability.** Move refreshes/embeddings to a replica; incremental MV strategies; HNSW/ivfflat tuning for vector scale.
16. **Security.** `ulda_web` read-only; `v_lead_llm_safe` strips direct PII before prompts; activation is consent-gated.
17. **Data governance.** Metric catalog (one governed definition per metric); PII classification; reverse-ETL audit; lineage from MV back to source.
18. **Performance.** Materialisation + covering indexes; `CONCURRENTLY` refresh; pre-warmed caches; read-replica offload.
19. **Future extensibility.** Real-time (streaming) Lead-360; ML scoring v2 via model registry; RAG over `lead_embedding`; new activation destinations.

Components (Domain D)

- **mv_lead_360** — *Materialised view*. The unified, flattened golden profile (identity + geo + completeness + sources). Exists for instant single-customer reads. Built on `golden_record_clean`; refreshed `CONCURRENTLY`. Apps render it; AI joins features to it. Example schema: `(master_lead_id uuid, name, first_name, last_name, company, email, phone, street, city, state, zip, completeness numeric, source_systems text[], source_count int)`.
 - **v_lead_journey_flat** — *Standard view*. Pivots `interaction` into `1st/2nd/3rd/4th channel + date + campaign` and `last channel/date` via window functions — your requested columns, computed not stored, uncapped. Apps/exports read it; analytics measures multi-touch paths. Materialise if it becomes hot.
 - **mv_lead_score** — *Materialised view*. Per-lead score (avg $\approx 76/100$) and tier (A/B/C/D) from a governed scoring function with MLOps lineage (`model_registry`, `score_run`), so an ML model can replace the heuristic with no consumer change. Apps prioritise by tier; AI consumes score as a feature.
 - **lead_embedding (pgvector)** — *Derived table + index*. 384-dim BAAI/bge-small vectors per lead with an HNSW index for cosine semantic search (`semantic_search()`). Exists to power similarity, dedup assistance, and RAG. AI uses it directly; the mass-embed runs on the replica/idle to protect the primary. `(master_lead_id uuid PK, embedding vector(384))` + HNSW.
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9. Cross-Cutting Capabilities

- **How it enables Lead-360.** One consumer object (`v_lead_master_flat`) joins `mv_lead_360` ⋈ `v_lead_journey_flat` ⋈ last-enrichment ⋈ last-order ⋈ serviceability → every requested column from one read, always correct.
 - **How it supports AI / semantic search / ML.** Clean per-domain **feature store**; `lead_embedding` + HNSW for semantic search; `v_lead_llm_safe` for prompt-safe context; `model_registry` for swapping heuristic scoring with trained models; append-only events are ideal ML training data.
 - **How it supports analytics & reporting.** Governed **metric catalog** (one definition per metric) + materialised rollups (`mv_analytics_by_state/tier/source` , KPI board) refreshed off the OLTP path.
 - **How it supports a CDP.** Persistent unified profile + identity resolution + consent-gated **reverse-ETL** activation to Warroom/SMS/dialer.
 - **Scalability, maintainability, performance.** Event volume scales independently of identity; partitioning + retention bound hot data; MVs isolate read load; schema boundaries make each domain independently evolvable and testable.
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10. Final Architectural Assessment

10.1 Why this is enterprise-grade

It exhibits the defining properties of enterprise data platforms: a single source of truth with immutable identity; complete lineage and auditability; separation of concerns by change-velocity; governed metrics and PII/consent controls; zero-downtime, reversible evolution; and independent scalability of identity vs activity vs serving. It is not a schema — it is a governed platform with an ingestion contract, a resolution engine, a serving layer, and a governance spine.

10.2 Vs traditional CRM database design

Traditional CRMs centre on a wide `Contact / Lead` table with bolted-on activity tables and app-enforced dedup. ULDA inverts this: identity is *resolved and mastered* (not assumed), activity is *event-sourced* (not last-value), serving is *materialised* (not live-joined on every page), and governance is *first-class* (not a spreadsheet). The result is better dedup, real history, cleaner AI, and a genuine path to a billion rows.

10.3 Recommended improvements before production-hardening

1. Make `interaction` & `enrichment` **partitioned append-only now** (they are empty — ideal timing).
2. **Backfill** `interaction` from `disposition_event` + Warroom SMS/dialer/reply activity — the one data task that lights up the journey/agent history.
3. **Stand up a read replica** to offload analytics/embeddings and unblock full 1 M embedding.
4. **Automate partition roll-out** (`pg_partman`) and retention.
5. **Formalise the consent gate as fail-closed** in every activation path (TCPA).
6. **Add a** `timezone` **attribute** (derived from state/ZIP) for quiet-hours dialing.

10.4 Additional domains/components worth adding at millions-of-leads scale

- **Suppression / compliance domain** (DNC registries, litigators, opt-outs) — beyond per-lead consent.
- **Household / account graph** (already seeded: `household` , `account`) — B2B and residential rollups.
- **Attribution / marketing-performance domain** (spend $\bowtie$ campaign $\bowtie$ conversion).
- **Data-quality & observability domain** (freshness SLAs, anomaly detection) — extend `ulda_gov` .
- **Real-time streaming tier** (CDC $\rightarrow$ bus) when sub-minute Lead-360 is required.

10.5 Verdict

The four-domain design is the correct enterprise architecture for this problem. Implemented on one PostgreSQL database with five schemas and a future read replica, it is scalable to millions of leads today and to a billion with the documented partitioning/replica path — while remaining maintainable, AI-ready, and production-safe. The primary remaining work is **data population (interaction backfill) and read scale-out (replica)**, not redesign.