# InsightDash

KPI Intelligence Platform

Owen

October 1, 2025

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### 1 Introduction

InsightDash is an interactive analytics application that aligns curated SQL datasets, reusable Plotly Dash components, and optional Large Language Model (LLM) insights into a single, leadership-ready command centre. The project ingests KPI extracts from a SQL Server instance, harmonises them across business lenses (outlet performance, regional comparisons, growth trends), and renders multi-tab dashboards. Each chart can be summarised via Gemini-powered narratives, giving business users quantified explanations without leaving the UI.

# 2 System Architecture

The runtime is orchestrated from the Dash entry point (app.py) and flows as follows:

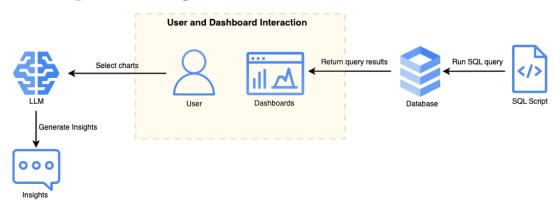
- 1. Configure structured logging and read environment-driven settings (database, LLM, feature switches).
- 2. Fetch per-tab datasets by executing templated SQL through SQLAlchemy engines.
- 3. Normalise and remap query outputs to match the figure contracts expected by each tab.
- 4. Initialise Dash layouts, set up shared colour palettes, and register interactive callbacks within app.py.
- 5. Serve interactive charts that support brushing, filtering, and multi-select synchronisation.
- 6. On demand, send chart data to Gemini for map-reduce style summarisation and cross-chart synthesis.
- 7. Persist usage logs and optional insight markdown while keeping the UI responsive.

### 2.1 Project Structure

	Dash bootstrapper and callback wiring
app_tabs	
	Outlet performance view
initpy	
figures.py	
layout.py	Tab layout components
tab2	Benchmarking scatter view
$\_\_$ init $\_$ .py	
figures.py	
layout.py	
tab3	Trend and ranking view
$\_$ initpy	
figures.py	
layout.py	
$igsqcup$ tab_compare	Outlet comparison tooling

```
figures.py
 layout.py
 settings.py ...... Environment flags and connection URIs
_logging.py .....Loguru sink setup
data_layer
_base.py ......Shared SQL execution helpers
 tab_1.py ......Tab-specific query orchestration
tab_2.py
tab_3.py
services
prompts.py ......Prompt builders for Gemini
sql_queries ...... Templated SQL per tab
utils
_df_summary.py .................Descriptive statistics for LLM payloads
```

### 2.2 Component Diagram



The primary components collaborate as follows:

- 1. Dash UI: Hosts the tabbed layout, callback graph, and controls.
- 2. **Data Layer**: Executes parameterised SQL templates and reshapes results for plotting.
- 3. **Services**: Provides LLM summarisation, prompt construction, and statistical preprocessing.
- 4. Users: Trigger filters, generate insights, and export findings.

5. External Systems: SQL Server supplies KPI tables; Gemini API enriches narratives when configured.

#### 2.3 Data Sources

InsightDash targets the cr\_kpi schema within SQL Server, expecting monthly tables (e.g., kpi\_april, kpi\_may) that expose outlet-level KPI metrics such as rate\_performance, rate\_quality, revenue shares, and intake volumes. Each data layer module wires its own SQL map:

- Tab 1: Outlet segmentation and region/category splits.
- Tab 2: Benchmark scatter plots with configurable axes and KPI filters.
- **Tab 3**: Time-sliced KPIs with ranking overlays.
- Compare Tab: Alias-based outlet lookup for side-by-side stat blocks.

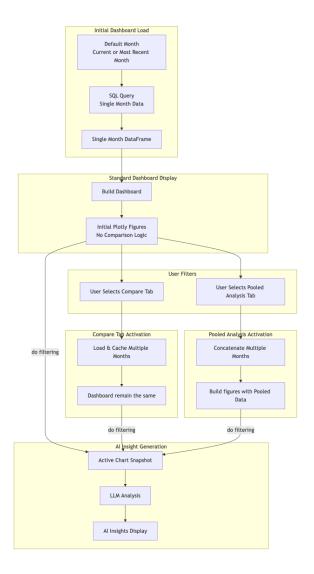
# 3 Implementation

### 3.1 Runtime Algorithm

The Dash application follows the control flow summarised in Algorithm 1. Each stage is idempotent, enabling hot-reload in development and reliable deployments.

```
Algorithm 1 InsightDash Runtime Lifecycle
Require: Environment variables, SQL templates, Plotly Dash runtime
Ensure: Responsive dashboard with optional LLM summaries
 1: CONFIGURE_LOGGING
 2: data_tab1 \leftarrow GET_TAB1_RESULTS("kpi_april")
 3: data_tab2 ← GET_TAB2_RESULTS("kpi_april")
 4: data_tab3 ← GET_TAB3_RESULTS("kpi_april")
 5: monthly ← LOAD_ADDITIONAL_MONTHS(["kpi_may", ...])
 6: app \leftarrow CREATE\_DASHBOARD(data\_tab1, data\_tab2, data\_tab3, monthly)
 7: while app is running do
       Process callback inputs and recompute filtered DataFrames
 8:
       if user requests insight then
 9:
          slices \leftarrow CHUNK_DATAFRAME(selection)
10:
          chunk_notes ← SUMMARIZE_CHART_VIA_CHUNKS(slices, context)
11:
          markup ← SYNTHESIZE_ACROSS_CHARTS(chunk_notes)
12:
          Update markdown panel with markup
13:
14:
       end if
       Append usage metrics to logs/usage.log
15:
16: end while
```

#### 3.2 Flow Chart



## 3.3 Key Modules

- app.py: Entry point that builds the Dash instance, injects tab layouts, and defines callback wiring for filters, graph selections, and insight generation.
- data\_layer/\*: Encapsulates SQL execution, result validation, and schema remediation so downstream charts always receive the expected columns.
- app\_tabs/\*/figures.py: Houses Plotly figure factories and table builders, isolating complex styling logic from callbacks.
- services/insights.py: Implements the map-reduce summarisation pipeline, including descriptive statistics to ground LLM outputs in verifiable numbers.
- utils/colors.py & utils/data.py: Provide consistent colour assignments, deduplication helpers, and DataFrame serialisation utilities shared across tabs.
- assets/\*.css: Defines the visual identity (typography, spacing, panel controls) to keep charts and sidebars visually cohesive.

# 4 Key Features and Advantages

### 4.1 Interactive Cross-Filtering

Linked controls propagate selection state across tabs, allowing users to drill into outlet types, regions, and KPI ranges without reloading the app.

### 4.2 Stable Visual Identity

Colour maps and plot templates are centrally managed, so repeated visits guarantee consistent legend ordering and palette usage, aiding comparative analysis.

#### 4.3 LLM-Assisted Narratives

When configured with a Gemini API key, business users can request per-chart insights or aggregated briefings that cite observed metrics and recommend next steps.

### 4.4 Operational Observability

Loguru sinks (stderr optional, file-based usage logs by default) capture database execution health, LLM token consumption, and user-triggered insight events.

### 5 Limitations and Risks

### 5.1 Database Connectivity

The dashboards rely on live SQL Server access. Network outages, credential drift, or schema changes will surface as empty charts despite defensive fallbacks.

# 5.2 LLM Configuration

LLM features degrade gracefully when GOOGLE\_API\_KEY is absent, but any misconfiguration (quota limits, model name changes) results in insight errors surfaced to the UI.

# 5.3 Data Schema Assumptions

Many figures expect specific KPI columns (e.g., rate\_performance, total\_score). Missing or renamed fields require corresponding updates to SQL templates and remap logic.

# 6 Related Technologies and References

- Plotly Dash documentation for component APIs and layout patterns.
- SQLAlchemy for engine creation and query execution against SQL Server.
- Google Gemini API for configuring the LLM insight features.
- Loguru for advanced logging patterns used by the project.

# 7 Frequently Asked Questions

#### • How do monthly KPI tables map to dashboard tabs?

Each tab module builds a SQL map using the configured month (e.g., kpi\_april). The keys (q1, q2, etc.) line up with figure factories that expect specific schemas.

#### • Can I run InsightDash without Gemini?

Yes. The LLM services detect missing credentials and return informative fallback text while leaving all visual analytics functional.

#### • What ensures colour consistency across tabs?

utils.colors.color\_map\_from\_list generates deterministic palette assignments based on sorted category lists, and the results are cached per session.

#### • Where are usage analytics stored?

Loguru writes structured entries to logs/usage.log, tagging events with usage=True so downstream tooling can aggregate adoption metrics.

#### • How do I add a new KPI tab?

Scaffold a new folder under app\_tabs, implement layout.py and figures.py, then wire new callbacks inside app.py (or a dedicated module) and extend data\_layer to supply the required SQL results.