

The background is a solid orange color. In the top left, there are two orange slices. In the top right, a hand is holding a glass dropper with a gold band, containing a yellow liquid. In the bottom left, there is another orange slice and a green mint leaf. A white line starts from the bottom right, goes up to the hand, then down to a point, and then left to a dot.

End-to-End Business Intelligence Case Study

# Titanium Dioxide Exposure & Regulatory Risk Intelligence **U.S. Cosmetics Industry**

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Business Intelligence | Data Analytics | Management Consulting

Power BI • SQL • Python • Data Modeling • DAX

# Executive Summary

## Problem

"Chemical exposure risk in the cosmetics industry is poorly quantified, despite increasing regulatory scrutiny."

## What I Built

A full end-to-end BI solution to quantify Titanium exposure across products, categories, and companies.

## Why It Matters

Any regulatory restriction would cause industry-wide disruption, not isolated impact.

## Key Result

- **86.6%** of cosmetic products contain Titanium
- **75%** of companies are exposed
- Largest brands carry the highest absolute risk
- Titanium Dioxide emerged as the dominant risk driver through analysis



# Business Context & Objective

## Industry Context

- Cosmetics rely heavily on chemical additives
- Titanium Dioxide used for pigmentation, opacity, and UV protection
- Increasing global regulatory and ESG pressure

## Objective

Answer four business questions:

1. How widespread is Titanium usage?
2. Which product categories are most dependent?
3. Which companies face the highest risk?
4. Where is regulatory exposure concentrated?

### Data Source

California Safe Cosmetics Program (CSCP)

### Data Scale

- 36,972 cosmetic products
- 604 companies
- 123 regulated chemicals
- Reporting period: 2009-2020

### Data Grain

Each record represents **one chemical used in one product**



# Dataset



# Data Preparation & Validation

## Steps Performed

- Raw CSV ingestion
- Python-based data audit and cleaning
- Standardized dates, chemical identifiers, and categories
- SQL-based KPI validation
- Cross-validated product counts and penetration metrics between Python, SQL, and Power BI to eliminate aggregation drift.
- Ensured consistency across Python, SQL, and Power BI

Why This Matters

Guarantees **trustworthy metrics** for business decision-making.



# Data Model (Star Schema)

## Why Star Schema

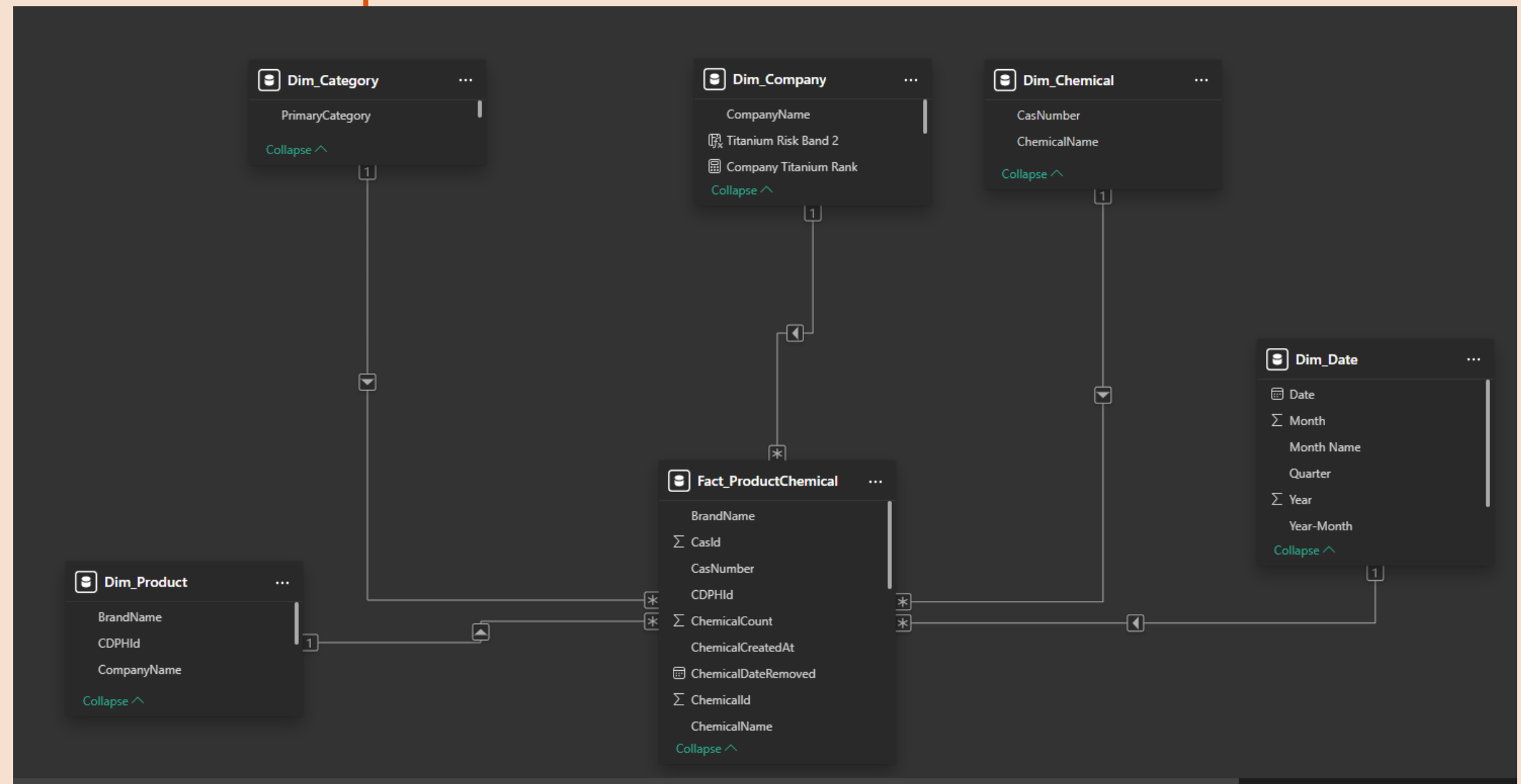
- Accurate aggregations
- Scalable BI design
- Reliable DAX calculations
- Supports time intelligence and drill-downs

## Fact Table

- Product x Chemical relationships

## Dimension Tables

- Product
- Company
- Chemical
- Category
- Date

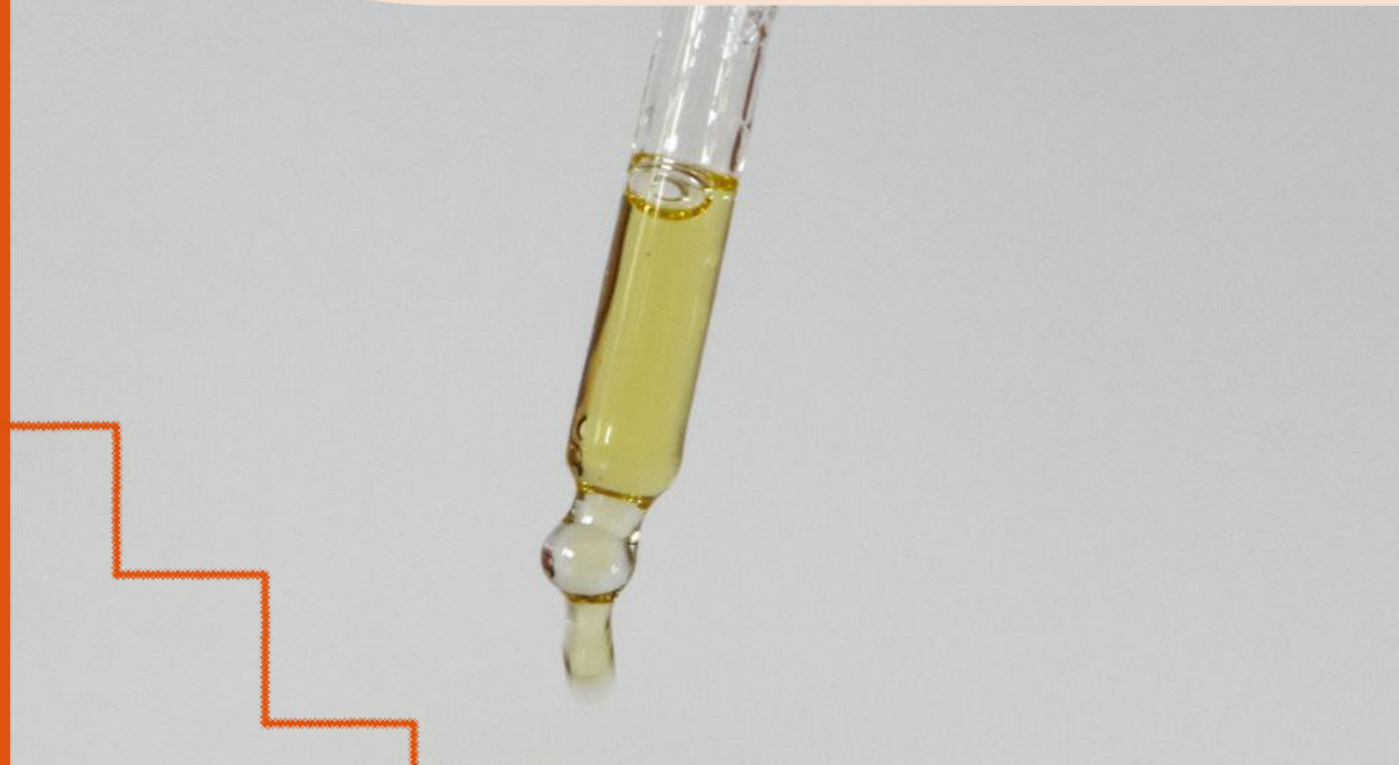


### Core KPIs

- Total Products
- Titanium Products
- Titanium Penetration %
- Companies Using Titanium

### Risk Innovation

"Percentage-based rankings masked true exposure; absolute Titanium product volume revealed systemic risk."



# Key Metrics & Logic



# Market-level Insights

## Findings

- 86.6% of cosmetic products contain Titanium
- Dependency remains high over 12 years
- No meaningful shift away from Titanium

## Business Interpretation

Titanium is a **structural dependency**, making regulatory action highly disruptive.

### Market-Level Titanium Exposure Overview

*"Analysis of product disclosures across U.S. cosmetics (2009–2020)"*

37K

Total Products

123

Total Chemicals

R  
I  
S  
K

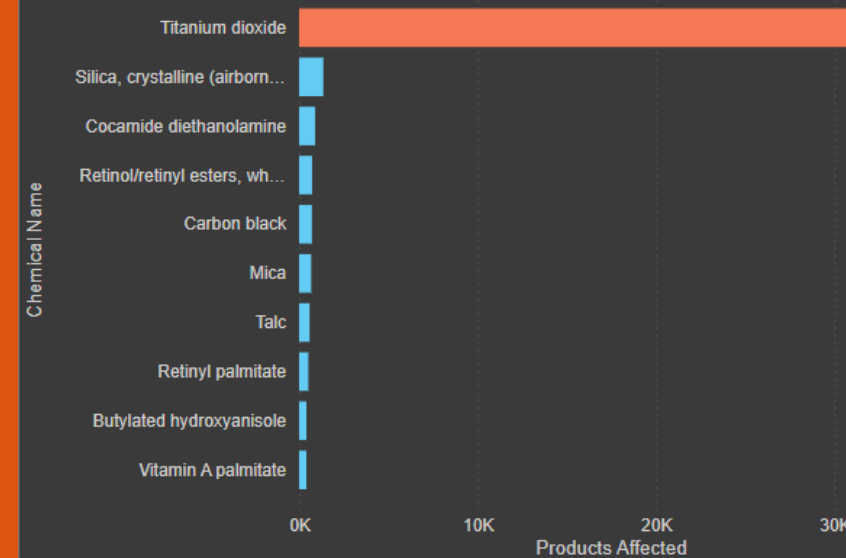
32K

Titanium Products

86.58%

Titanium Penetration %

#### Top Chemicals by Product Exposure



#### "Filter by Category"

PrimaryCategory

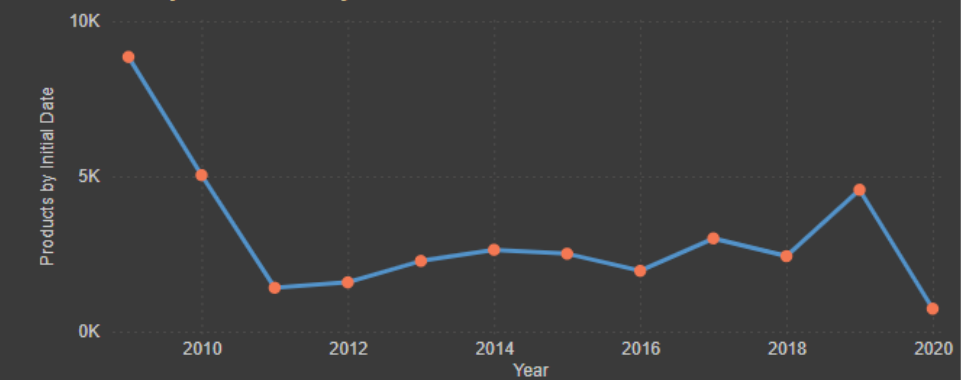
All

#### "Filter by Company"

CompanyName

All

#### Products by Initial Date by Year





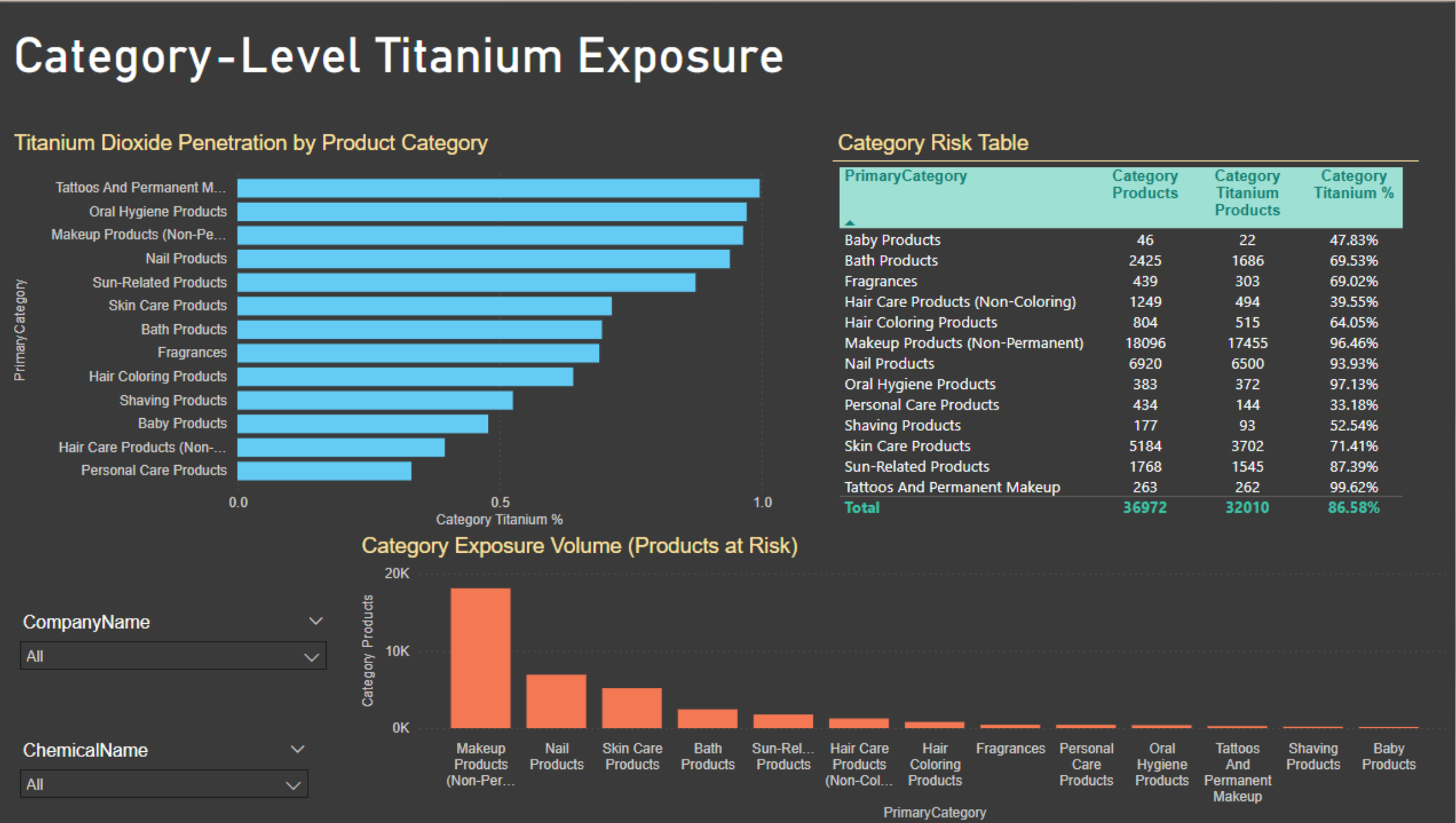
# Category-level Risk Insights

## Highest Risk Categories

- Makeup (Non-Permanent)
- Nail Products
- Oral Hygiene
- Tattoos & Permanent Makeup

## Key Insight

These categories are **chemically locked-in**, making reformulation costly and complex.



# Company-level Risk Insights

## Risk Concentration

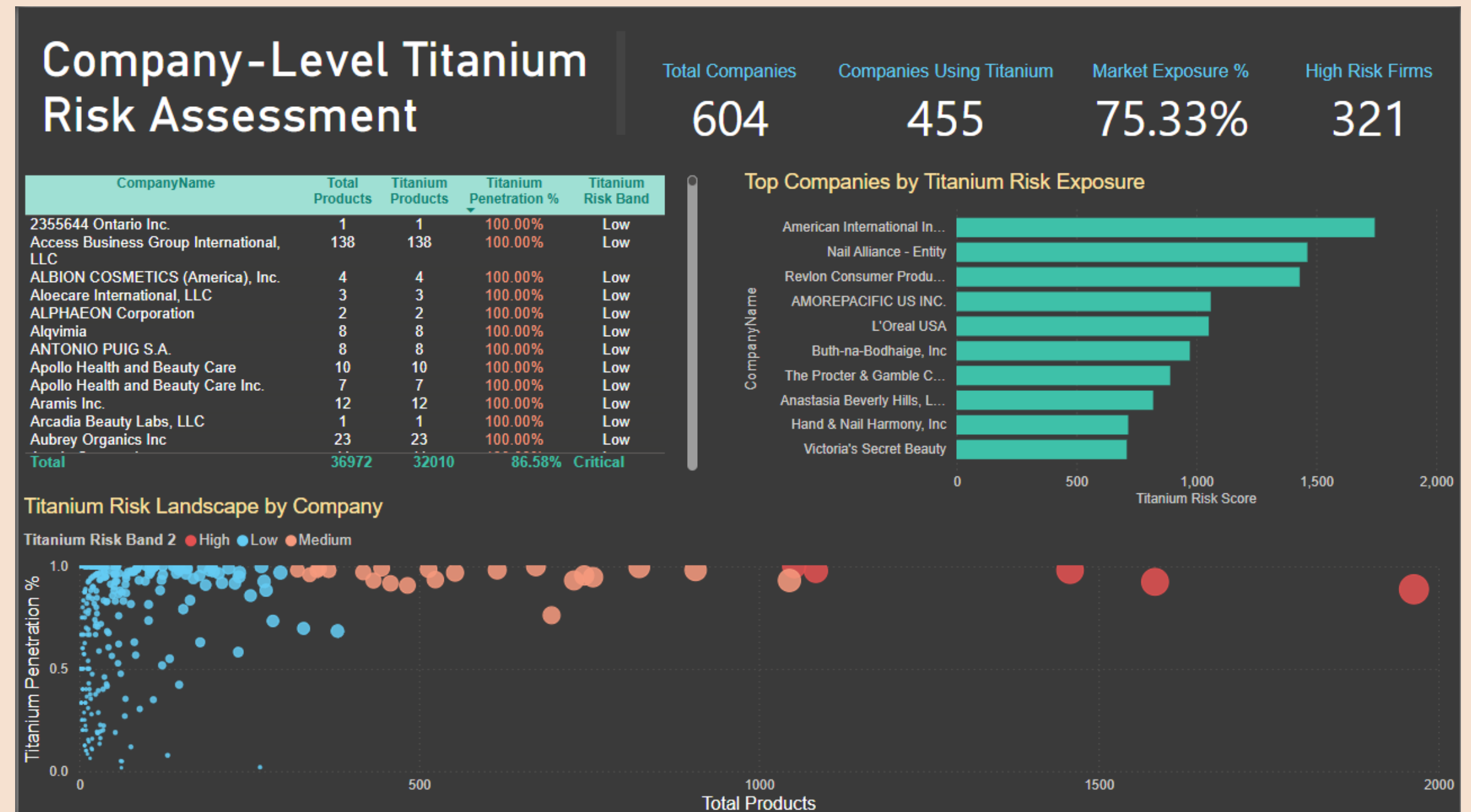
- Large companies dominate Titanium exposure
- High dependency x large portfolios = extreme risk

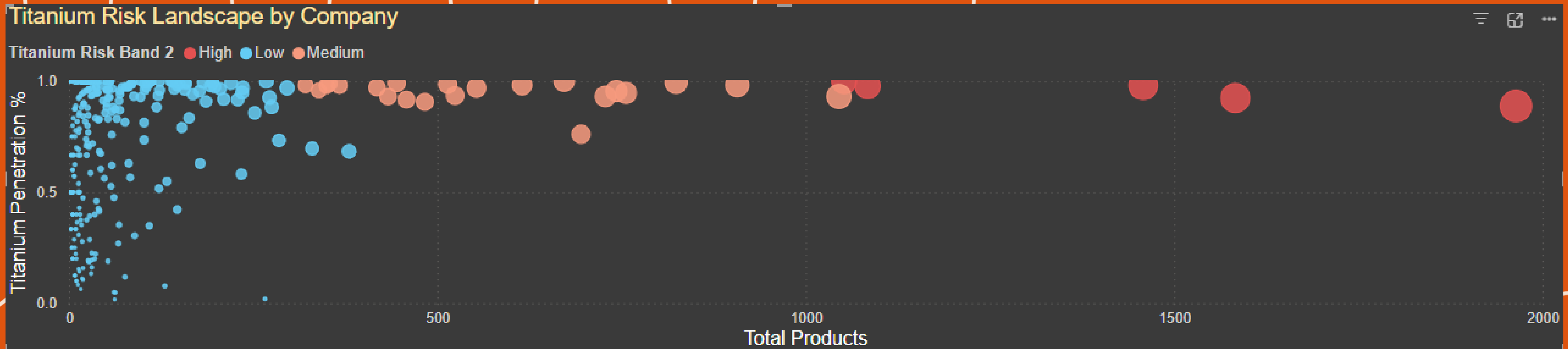
## Top Exposure Companies

- American International Industries
- Nail Alliance
- Revlon
- AMOREPACIFIC
- L'Oréal USA

## Insight

Regulatory action would impact **market leaders first**, not fringe brands.





# Risk Landscape (Scatter Analysis)

## What the Chart Shows

- X-axis: Company size (Total Products)
- Y-axis: Titanium dependency (%)
- Bubble size: Titanium exposure volume
- Color: Risk band (Low → Critical)

## Why This Matters

Visually identifies **systemic risk clusters** in the industry.

# Business & Regulatory Implications



## **For Regulators**

Focus on high-impact companies and categories first

## **For Companies**

- Prioritize reformulation in high-risk segments
- Diversify ingredient sourcing

## **For Investors**

"Titanium exposure represents an unpriced ESG and regulatory risk that could materially impact large-cap cosmetics portfolios."



# Skills Demonstrated



## What This Project Proves

- End-to-end BI ownership
- Strong data modeling fundamentals
- Advanced DAX for ranking, context control, and risk banding.
- Business-driven insight generation
- Dashboard design for decision-makers

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A background image showing a person's profile as they drink from a bottle. The bottle has a label that says "TONIC". The image is faded and serves as a backdrop for the text.

"Regulatory risk is structurally concentrated  
– **driven by product design**, not isolated  
compliance failures."





# Thank you

This project demonstrates how bi can be used to **anticipate risk**, not just report numbers.