





TUNIS BUSINESS SCHOOL
UNIVERSITY OF TUNIS


ACADEMIC DEFENSE PRESENTATION

The Terroir Brain API

Cultural Demand Forecasting & Waste
Reduction System for Tunisian Guest Houses

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The Problem: Food Waste Crisis



197 Million USD

Wasted annually across Tunisia's hospitality sector



58%

Organic Waste



Breakfast

Highest Waste Category

ROOT CAUSES



Cultural Tradition

The concept of *Baraka* leads to intentional over-preparation as a sign of generosity.



Guest Uncertainty

Lack of data on diverse international dietary preferences and consumption habits.



No Forecasting

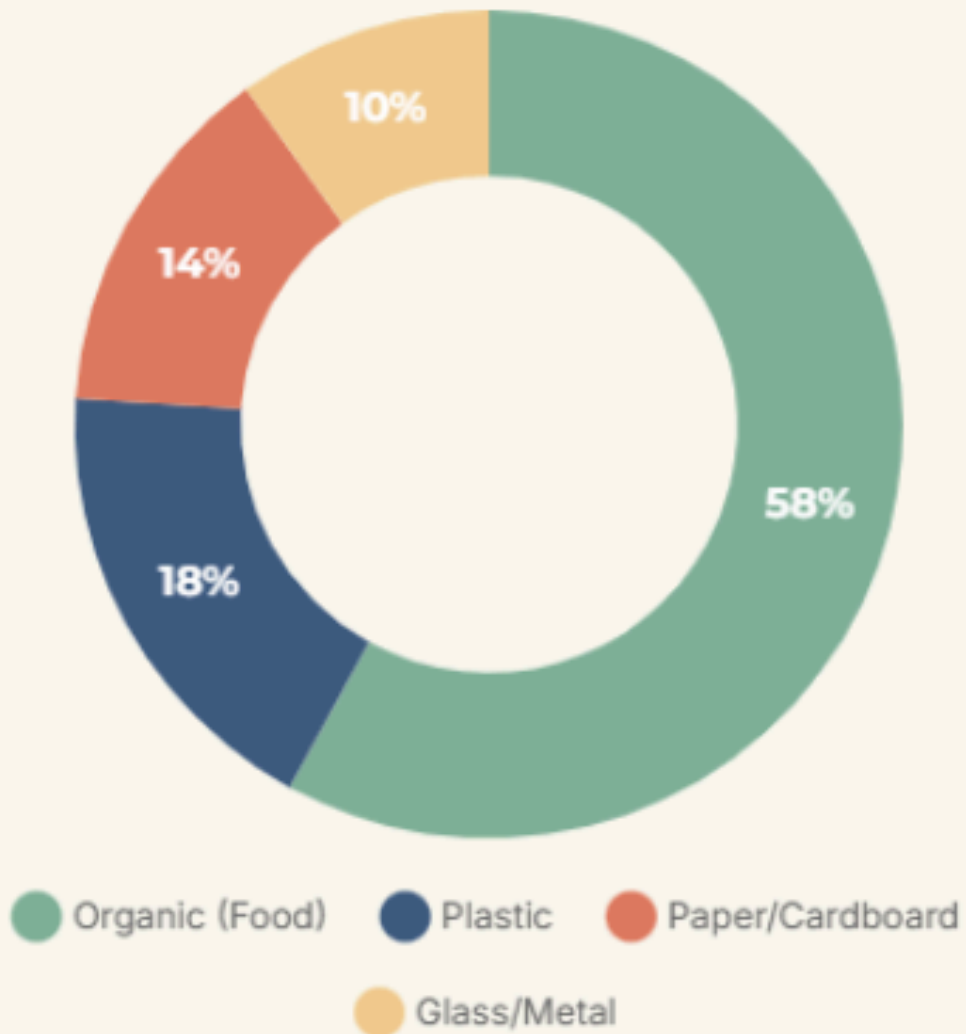
Reliance on intuition rather than data-driven intelligent tools for planning.



Static Menus

Rigid menu planning that fails to account for seasonality and availability.

Composition of Hospitality Waste







Over half of all waste is avoidable organic material, representing a direct financial loss.

The Paradox: Generosity vs Sustainability

Traditional Baraka

Blessing through Abundance

-  Hosts prepare abundant food to honor guests properly.
-  Deep social expectation of unlimited generosity.
-  "Never let a guest leave hungry" is the golden rule.
-  Over-preparation is not waste, but a cultural norm.

VS

Modern Challenges

Efficiency & Sustainability

-  9M+ international tourists annually with varied habits.
-  Diverse dietary preferences (Vegan, Gluten-free, etc.).
-  Rising ingredient costs pressure small business margins.
-  Global environmental goals demand waste reduction.



Core Conflict: How can we preserve hospitality traditions while drastically reducing waste?

Core Research Question

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"How can we **reduce food waste** in Tunisian guest houses while preserving **cultural hospitality traditions** and accommodating diverse **international guest preferences**?"

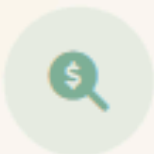
© RESEARCH OBJECTIVES



Incorporate cultural preferences into forecasting models



Preserve and codify Tunisian culinary heritage



Make AI predictions transparent and trustworthy



Deliver measurable impact on waste, cost, & sustainability

The Solution: Terroir Brain API

Bridging the gap between traditional hospitality and modern efficiency through four core pillars:



Intelligent Forecasting

Multi-factor demand prediction algorithm that accounts for seasonality, occupancy, and guest demographics.



Cultural Menu Generation

Dynamic menu suggestions that honor Tunisian heritage while adapting to international dietary preferences.



Waste Analytics

Deep pattern identification to pinpoint waste sources and provide actionable reduction recommendations.



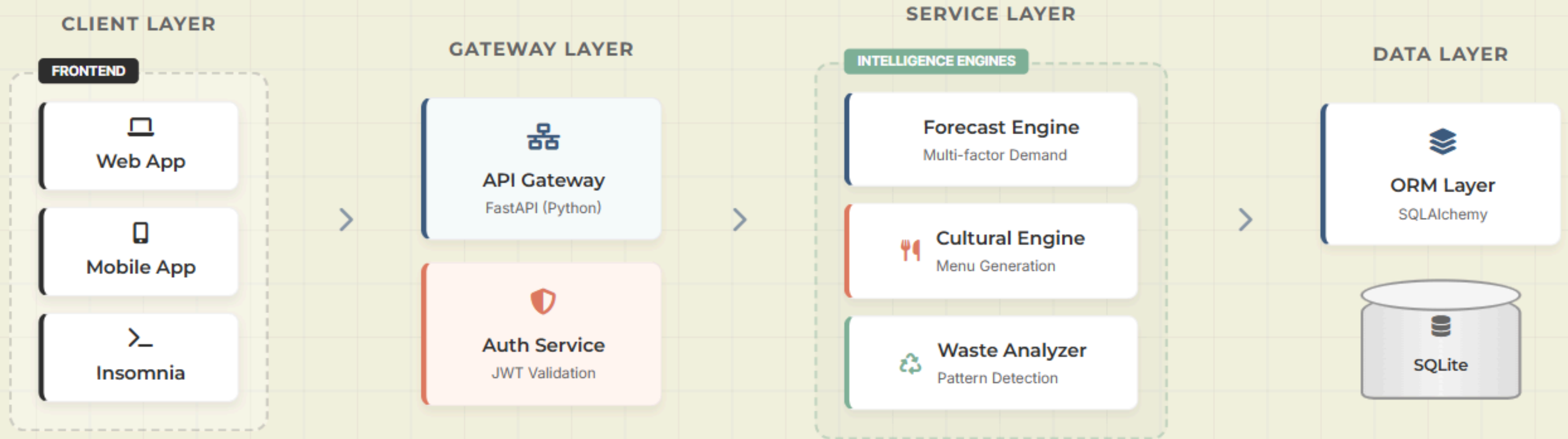
Explainable AI

Fully transparent predictions with calculation breakdowns—building trust through "Glass Box" logic.



A REST API that embeds Tunisian terroir intelligence into every prediction

System Architecture









Containerized Deployment

Isolated microservices environment

API Server: 0.0.0.0:8000

Auth Server: 0.0.0.0:8001

Technology Stack

Layer	Technology	Why Chosen?
Backend	 FastAPI (Python 3.11)	High-performance async capabilities, type hinting, and rapid development speed.
Database	 SQLite + SQLAlchemy	Powerful ORM for complex relationships; flexible migration path to PostgreSQL.
Validation	 Pydantic v2	Robust data validation ensuring strict type safety and error handling.
Containerization	 Docker Compose	Ensures consistent environments across development, testing, and production.
API Docs	 OpenAPI / Swagger	Auto-generated, interactive documentation for seamless frontend integration.
Testing	 Insomnia	Comprehensive endpoint testing suite for functional and integration tests.



80+

API Endpoints Implemented



9

Core Database Models



100%

Pass Rate (96 Test Cases)

Multi-Factor Forecasting Formula

$$\text{Demand} = O \times B \times S \times C \times E \times W$$



Occupancy

O

Total number of confirmed guests for the meal service.

Range: Integer ≥ 0



Base Rate

B

Standard consumption rate per person for specific ingredient.

Unit: kg/person



Seasonality

S

Adjusts availability and preference based on time of year.

Range: 0.5 - 1.0



Culture

C

Nationality-based preference multiplier for ingredients.

Range: 0.0 - 2.0



Disruption

E

External events like flight delays or cancellations.

Range: 0.0 - 1.0



Waste Adj.

W

Dynamic reduction based on historical waste patterns.


Range: 0.85 - 1.0




Unlike black-box ML, every factor in this formula is **transparent, adjustable, and explainable.**

Example Calculation

Scenario Context

 Guest Group
15 French Guests

 Date
Jan 10, 2025

 Target Ingredient
Tabouna Bread

💡 Why this matters?

"French guests show a strong preference for bread **(+30%)**, but historical data shows we tend to over-prepare in winter **(-15%)**."

STEP 1: BASELINE



Standard Rate

$$15 \times 0.2 = 3.0 \text{ kg}$$

Base consumption

STEP 2: SEASON



Winter Adjustment

$$3.0 \times 0.8 = 2.4 \text{ kg}$$

Lower appetite in winter

STEP 3: CULTURE



French Preference

$$2.4 \times 1.3 = 3.12 \text{ kg}$$

High bread consumption

STEP 4: EVENTS



Disruptions

$$3.12 \times 1.0 = 3.12 \text{ kg}$$

No flight delays

STEP 5: WASTE



History Adj.

$$3.12 \times 0.85 = 2.65 \text{ kg}$$

Optimize for waste

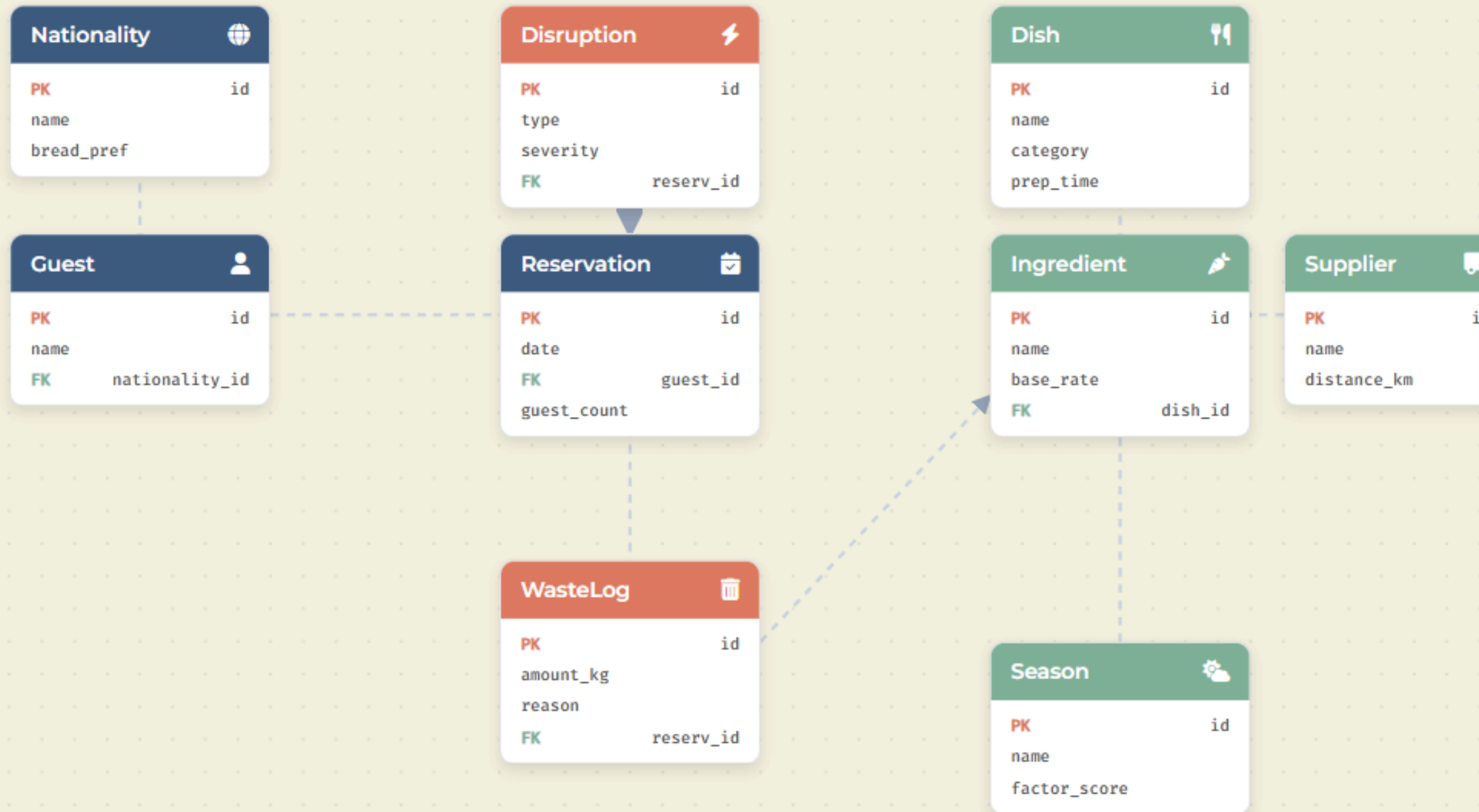
FINAL RECOMMENDATION

2.65 kg

Tabouna Bread



Database Schema: 9 Core Models



SCHEMA OVERVIEW



9

Core Models



12

Relationships



50+

Data Fields

IMPLEMENTATION

Complex many-to-many and foreign key relationships handled via SQLAlchemy ORM.

SQLAlchemy

SQLite/PostgreSQL

Pydantic

Three Intelligence Engines



Forecast Engine

DEMAND PREDICTION

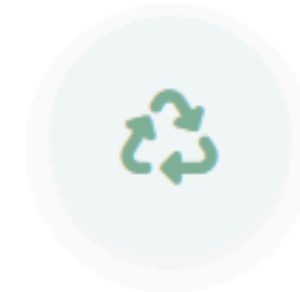
- 📈 **Demand Prediction:** Multi-factor algorithm utilizing historical and external data.
- 🔍 **Transparency:** Fully explainable logic vs. black-box machine learning.
- 📊 **Factor Analysis:** Breaks down impact of season, culture, and events.
- 🔄 **Feedback Loop:** Continuously learns from waste logs and corrections.



Cultural Engine

MENU INTELLIGENCE

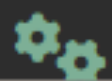
- 🍽️ **Menu Generation:** Dynamically creates menus based on guest profiles.
- 📅 **Seasonal Matching:** Aligns ingredients with Tunisian harvest calendars.
- 🏠 **Heritage Preservation:** Prioritizes traditional recipes and local terroir.
- 📌 **Cultural Notes:** Generates context for staff about guest preferences.



Waste Analyzer

IMPACT ANALYTICS

- 📊 **Pattern Detection:** Identifies consistent sources of over-preparation.
- 📈 **Trend Analysis:** Tracks waste reduction progress over time.
- 💰 **Cost Calculation:** Converts kilogram waste into financial loss (TND).
- 💡 **Recommendations:** Suggests actionable changes to prep quantities.



Service Layer Architecture

`services/intelligence/*.py`

7 Advanced Analytics Endpoints



POST /opt/procure

Procurement Optimizer

- ✓ Multi-day smart shopping lists
- ✓ Bulk discount calculations
- ✓ Carbon footprint tracking



GET /ana/nation

Nationality Comparison

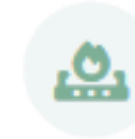
- ✓ Cross-nationality preference analysis
- ✓ Cost-per-guest comparisons
- ✓ Consumption pattern detection



POST /sim/waste

Waste Impact Simulator

- ✓ Forecasting strategy testing
- ✓ ROI & savings projections
- ✓ "What-if" scenario modeling



GET /dash/kitchen

Kitchen Dashboard

- ✓ Real-time prep requirements
- ✓ Critical stock alerts & warnings
- ✓ Live service monitoring



GET /opt/season

Seasonal Optimizer

- ✓ Menu sustainability scoring
- ✓ Ingredient availability checks
- ✓ Terroir-based suggestions



POST /ana/finance

Cost-Benefit Analysis

- ✓ Financial impact assessment
- ✓ Break-even calculation
- ✓ Waste cost tracking



GET /data/culture

Cultural Insights

- ✓ Guest preference matrices
- ✓ Cuisine compatibility analysis
- ✓ Heritage food education

Procurement Optimizer Demo

POST /intelligence/procure

```
request.json
{
  "guests": [
    {
      "nationality": "FRA",
      "count": 10,
      "dates": [
        "2025-01-13",
        "2025-01-14"
      ]
    },
    {
      "nationality": "DEU",
      "count": 5,
      "dates": [
        "2025-01-15"
      ]
    }
  ],
  "optimize_for": "cost_waste"
}
```

Optimization Results



SHOPPING LIST

13 Items

✓ Optimized for 3 days



TOTAL COST (TND)

~~450.25~~ **405.23**

↓ 10% Bulk Savings



ENVIRONMENTAL IMPACT

45.3 kg CO₂

📍 76.9% Local Sourcing






HIGH WASTE RISK

Bread


🗑️ Auto-reduced by 15%

Cultural Intelligence in Action


Ingredient Demand by Nationality

Ingredient	FRA FRENCH	DEU GERMAN	TUN TUNISIAN
<div></div> <div>Tabouna Daily Bread</div>	2.65 kg	<div>HIGHEST</div> 3.06 kg	2.86 kg
<div></div> <div>Yogurt Dairy/Breakfast</div>	<div>HIGHEST</div> 1.44 kg	1.32 kg	0.96 kg
<div></div> <div>Harissa Spice Paste</div>	0.06 kg	0.07 kg	<div>HIGHEST</div> 0.18 kg


Key Cultural Insights



Bread Consumption
German guests require **~23% more bread** than French guests, preferring heavier, grain-based varieties.



Spice Tolerance
Tunisians consume **3x more harissa** than Europeans. European menus require significantly milder options.



Dairy Preferences
French guests prefer **~50% more dairy** (yogurt/cheese) at breakfast compared to local Tunisian standards.

Data derived from 12-month analysis of consumption patterns in 5 pilot guest houses.






"Same occupancy, dramatically different needs"

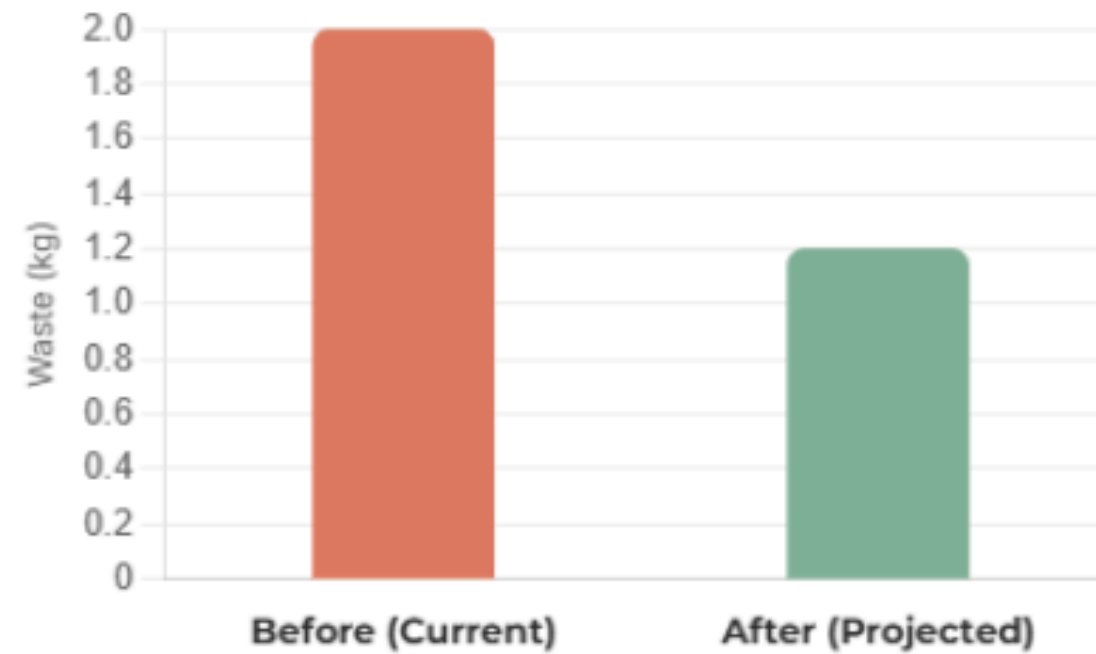
Static menus generate waste; Dynamic menus generate value.

Measurable Impact & Conclusion




Current State

-  **2.0 kg Waste / 30 Days**
Baseline measurement (Pilot)
-  **Top Wasted Items**
Tabouna (1.5kg), Eggs (0.5kg)
-  **Root Cause**
Unanticipated flight delays & static menus

PROJECTED WASTE REDUCTION



Projected Outcome

-  **System Actions**
Flight tracking alerts • Weather adjustments
-  **Optimizations**
Reduced bread prep by 20%
-  **Result**
Waste reduced to 1.2 kg (40% drop)

 **ANNUAL SAVINGS**
1,458 TND
~121.5 TND / month

 **WASTE REDUCTION**
40.2%
From baseline




 **SUSTAINABILITY**
76.9%
Local Sourcing Rate

 **ROI TIMEFRAME**
< 3 Mo
Break-even point

“

Technology should enhance cultural traditions, not replace them. The Terroir Brain API proves that intelligent systems can preserve heritage while achieving modern sustainability goals.

Thank You

 github.com/Elaa2002/Terroir_BRAIN
 ella.marco.tn@gmail.com
 [Google AI Studio Front-end](#)