



Smart UI/UX, Automation & Al-driven Insights for Asset Management Platform

**CASE STUDY** 

MANTRA LABS



Founded in **2003**, Viteos has been named **Best Shadow Accounting Firm 2019** for the sixth consecutive year, and was recently acquired by Capital Firm, **Intertrust**.



A **Technology Platform** offering smart solutions for Investment & Asset Management ranging from shadow-accounting to fund administration, for nearly every asset class in the World.

Currently, Viteos manages **\$350 billion** worth of assets globally.



The **financial asset management platform** operates a global delivery model that provides end-to-end middle & back office administration for top-tier hedge funds, private equity, private debt and other alternative asset managers.

Their platform was mired in several operational bottlenecks. **Mantra Labs** answered the call to solve some of their most pressing pain-points.

#### **USER JOURNEY**

The User's journey (UX/UI)

and routine workflows were not intuitive, creating difficulties in seamless access to data & reporting.

#### **Automated Onboarding**

A manual onboarding process to gather client details & request regulatory documentation slowed down client setup-times.

#### **Al-driven Insights**

Advanced machine learning algorithms to find the lowest-commission broker for a specific trade; extract faster data insights from historical P&L data.

### **Sol**ution

Mantra Labs refined their user workflows for more robust capabilities and smarter gains; integrated an automated client onboarding solution and created a machine learning model for producing actionable data insights.

### M

#### **Smarter Workflows**

Single screen interface, reduced cognitive load, intuitive UX, coherent navigation, and responsive dashboards.

#### **Automated Onboarding**

Historical data mapping, data reconciliation with source records, data cleansing and validation to auto-fill fields in client portal.

#### **Al-driven Insights**

Advanced machine learning algorithms to find the lowest-commission broker for a specific trade; extract faster data insights from historical P&L data.

## VuRec UX/UI

#### **OBSOLETE WORKFLOW & UX DESIGN**

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VuRec UX/UI

#### **REMODELED WORKFLOW & UX DESIGN**

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#### OLDER DASHBOARD VISUALIZATIONS

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VuRec UX/UI

## VuRec UX/UI



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### CLIENT ONBOARDING

Previously, when a new client was being onboarded – the client data, which can be in a variety of formats and taxonomy, was manually converted to a standard Viteos format – **causing 'onboarding fatigue**'.

#### **SmartETL** AUTOMATED ONBOARDING SOLUTION

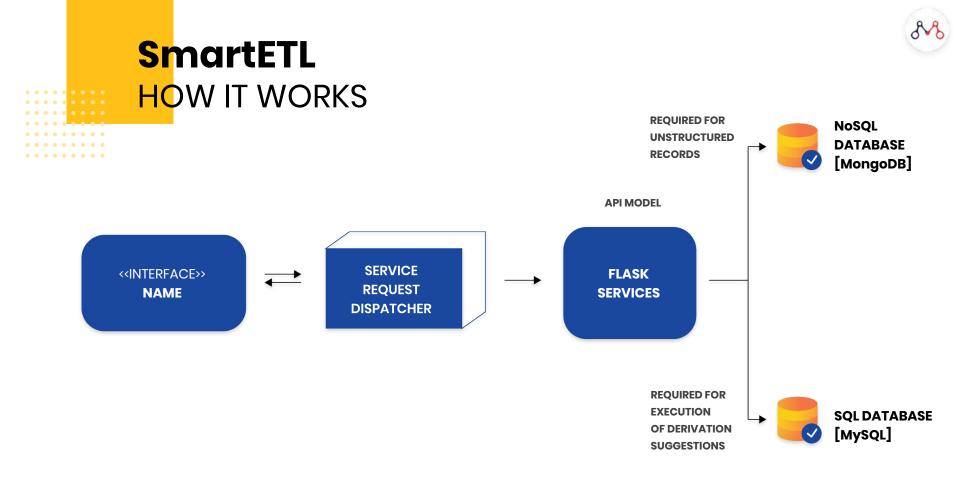
The SmartETL system maps historical records

and new client information in order to suggest faster and accurate inputs to the user.

#### **Key features**

- No restrictions on data source, file format, and taxonomy.
- Data reconciliation with source records.
- Data cleansing & validation, for auto-filling.

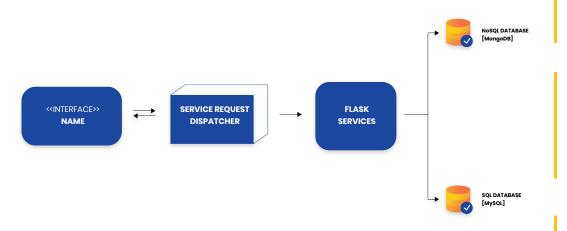




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### SmartETL HOW IT WORKS Contd...



#### **DATA EXTRACTION & TRANSFORMATION**

The system uses **NoSQL** & **SQL Server** databases to unite historical & modern data. Historical data is transformed into a NoSQL compatible format for model services. The SQL Server is used for executing certain SQL queries.

#### **API MODELLING**

Using **Python's Flask Framework** and **Pandas library**, the API takes user input and interacts with the database layer to view/update data.

#### **INTERFACE DESIGN**

The system is programmed in **Angular/React framework** and is independent of view layer i.e. services are accessible on web/desktop/mobile.



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#### **MACHINE LEARNING ALGO**

Earlier, data from multiple sources and distinct formats made it challenging for the user to comprehend historical data.

Using ML algorithms, large volumes of comprehensive data can be ingested & optimized for rapid analysis and prediction, while also simplifying time intensive documentation in data-entry.

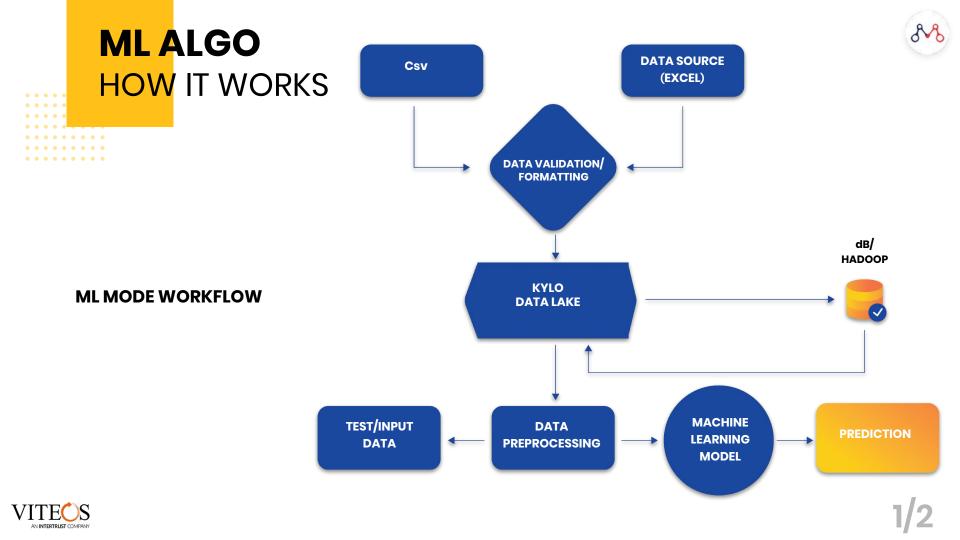
#### Al-Enabled Insights MACHINE LEARNING MODELS

The three ML models – **regression**, **outlier detection**, and **feature selection** analyse historical transactions, trades, and financial data from clients, accounting systems & banks.

#### **Key features**

- 1 Clients can find the lowest-commission broker for a particular trade.
- 2 Accurate and faster data insights.
- 3 Flexibility to use the models on different data sets.
- 4 Efficient memory utilization for large volumes of data.





### ML ALGO HOW IT WORKS Contd...

#### **MACHINE LEARNING MODEL PROCESS**

- 1 Fetching data from Kylo Data Lake
- 2 Preprocessing the data
- **3** Handling missing and null values
- 4 Perform dimensionality reduction
- 5 Data visualisation
- 6 Creating ML model
- 7 Performing validation
- 8 Checking for Model accuracy
- 9 Saving the Model

#### **APPLICATION & MODULES**

Regression: **Polynomial Regression** Outliers: **Isolation Forest** Feature Selection: **Light GBM** 

#### **END-USERS**

The results of all three models can be utilized by users at all levels including **CXOs** and **clients**.

#### **TECHNOLOGIES USED**

Python / Pandas / Numpy / Matplotlib / Plotly / Pyhive Light GBM / Seaborn / Scipy / Pyspark / Elasticsearch Scikit Learn / ActiveMQ / Apache NiFi / MySQL





### Key results



#### SmartETL CLIENT ONBOARDING SOLUTION

#### **REDUCED ONBOARDING TIME**

The user can onboard a client within 1-2 hours resulting in 87.5% efficiency gain.



#### **AI-DRIVEN INSIGHTS**

#### **FASTER DATA INSIGHTS**

Viteos' Clients (traders) save anywhere between **46-92%** in trade-brokerage.



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### **About** Mantra Labs

- Real World Problem Solving using Artificial Intelligence and Customer Experience Consulting
- Domain Focused Products and Solutions for the Digital Insurer

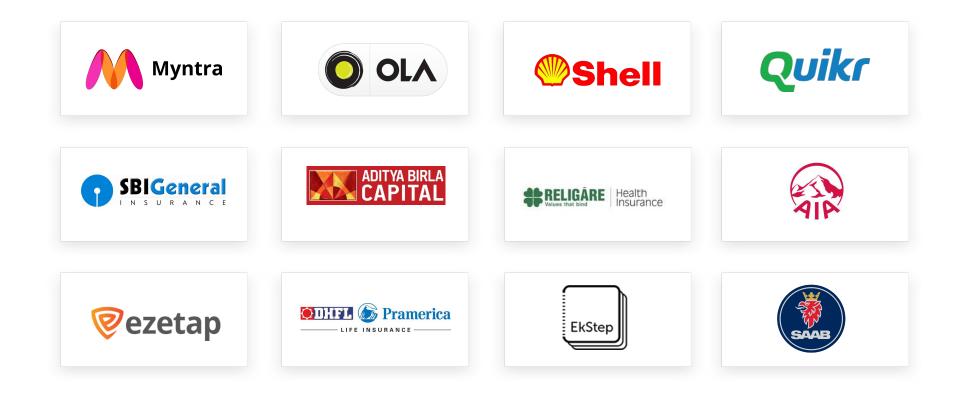


100+	150+	03
PROJECTS	GEEKS	OFFICES

#### **Key Partnerships**



## Key Clients



# Let's build what you need, together.

Get in touch with <u>us</u>.

