

## **Project Overview**

- Project Name: Robo Advisor Portfolio Recommender
- Goal: Build a web-based robo advisor that recommends personalized investment portfolios using modern portfolio theory (Markowitz Model).
- **Key Feature:** Dynamic portfolio recommendations based on the user's risk profile (Conservative / Balanced / Aggressive).

## **Problem Statement**

- **Problem:** Retail beginner-advanced investors struggle to build efficient portfolios aligned with their risk preferences.
- Challenges Faced by Users:
  - 1. Lack of financial knowledge
  - 2. Overwhelmed by stock market options
  - 3. No personalized financial guidance
- Need: A system that simplifies investment and provides trustable smart recommendations.

## Target Users & Value Proposition

## Target Audience:

- Beginner to intermediate investors
- Young professionals with limited financial experience

#### Value Provided:

- Data-driven portfolio suggestions
- Simple, intuitive interface
- Clear risk-adjusted trade-offs

## High-Level Architecture

## Frontend (Flask Web App):

- Built with Python + Flask, served over HTTP API.
- Allows users to select risk profile and view personalized portfolio
- Additional tabs: market statistics, investment view, leaderboard

## **Backend (Python):**

- Fetches financial data using yfinance (Yahoo Finance API)
- Core logic based on Markowitz Efficient Frontier model
- Portfolio simulation, optimization, and recommendation

## Backend Implementation (Tal Koskas & Ron Alima)

#### MarkowitzModel Class:

- Calculates expected returns and covariance matrix from historical prices
- Simulates 100,000 portfolios
- Computes Sharpe Ratio, Returns and Volatility

## ModelManager Class:

- Trains and stores models (e.g. "markowitz")
- Returns optimal portfolio per risk profile

# **Backend Implementation (Continued)**

## **Risk Profiles Implemented:**

- Conservative lowest volatility
- Balanced max Sharpe ratio
- Aggressive highest return

## **REST API Endpoint:**

 /api/portfolio: Accepts risk profile, returns recommended portfolio in JSON

## Smart Logic - Markowitz Model

- Inputs: Historical prices of selected tickers (2020–2025)
- Simulation:
  - Portfolio weights sampled randomly
  - Return, volatility, Sharpe calculated per portfolio
  - Efficient frontier plotted
- Selection: Best-fit portfolio for each risk profile strategy

## Frontend Implementation (Ron Alima)

- Built using Flask + HTML/CSS/JS
- UI Elements:
  - Dropdown for selecting risk profile
  - Portfolio display: ticker + allocation + return
  - Additional tabs: Statistics, Investments, Leaderboard, Profile
- Profile Tab:
  - User can input budget and view share profit
- Statistics Tab:
  - Shows user budget
  - Graph of SPY share performance over time (day/week/month)

# Frontend Implementation (Continued)

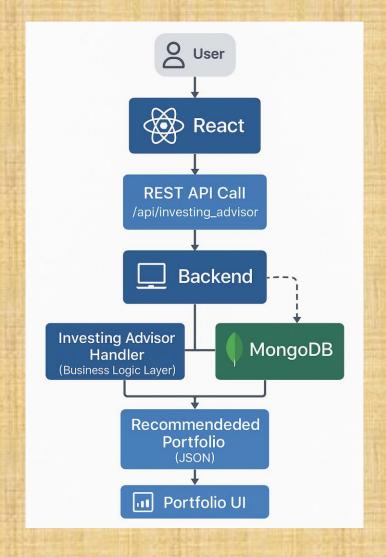
#### Investments Tab:

- Displays budget, shares profit, achievements
- "My Positions" table: symbol, shares, entry price, type, % and \$
  difference, stop limit, add new position option

#### Leaderboard Tab:

- Displays achievements (points), rank, shares profit
- · Ranks table: rank, name, achievements, profit
- Uses AJAX to call Flask API
- Responsive design for easy use on desktop/mobile

# Architecture Flow – Portfolio Recommendation Flow



## User Interface Features

- - Login/Register Dialog (MUI Dialog + Validation)
- - Portfolio Page: Dynamic allocation table per user
- Statistics Tab: SPY chart + budget display
- - Investments Tab: Positions, achievements, gain/loss
- - Leaderboard Tab: AJAX-based real-time updates

## Backend Architecture Highlights

- - Built with ASP.NET Core 8 (C#)
- - Controllers (e.g., InvestingAdvisorController)
- - Business Logic Layer: Service handlers per domain
- - MongoDB integration (Users, Positions, Risk Profiles)
- REST API: returns recommended portfolio integrated with tal's part.
- - Coordination with Python ML model via internal service

## Demo Time - Live Use Case

- Input: User selects risk profile (via dropdown)
- Output:
  - Portfolio recommendation: e.g., SPY: 20%, AAPL: 15%, etc.
  - Visualization of efficient frontier
  - Interactive tab interface

## Challenges & Future Plans

#### · Challenges:

- Financial data accuracy and availability
- High computational load for simulations
- User-friendly interface for non-technical users

## Next Steps:

- Add LSTM-based return forecasting module (deep learning)
- Integrate database (user sessions, portfolio history)
- Deploy full system to cloud (e.g., AWS/Heroku)
- Improve interactivity with real-time refresh and analytics

# Summary & Q&A

#### Progress:

- End-to-end MVP: backend logic, API, and frontend interface
- · Real data, real logic, real-time demo

## Learning Outcomes:

- Full-stack system implementation
- Practical use of finance + ML concepts
- Collaboration across backend and frontend
- Any questions or feedback?