# **Contextual Bandit Portfolio Allocation**

Evaluating LinUCB contextual bandit for weekly capital allocation across five technology leaders: AVGO, AAPL, TSLA, GOOGL, NVDA.

### **Test Period**

52-week evaluation window from September 23, 2024 to September 22, 2025

#### **Context Features**

Lagged returns, volatility, RSI, and relative price/volume signals drive allocation decisions

## Benchmarks

Equal-weight portfolio, seeded random arm selection, SPY ETF, and best single ticker performance



# LinUCB Architecture Overview

LinUCB feedback loop: feature engineering  $\rightarrow$  state update  $\rightarrow$  UCB policy  $\rightarrow$  realized reward feedback



### **Feature Engineering**

Technical indicators and market signals transformed into context vectors for each asset



## State Update

Bayesian parameter estimation updates confidence bounds based on observed rewards



#### **UCB Selection**

Upper confidence bound policy balances exploitation of high-performing arms with exploration



#### **Reward Feedback**

Weekly returns provide reward signal to refine future allocation decisions

## **Performance Results**



90.9%

2.06

-26.7%

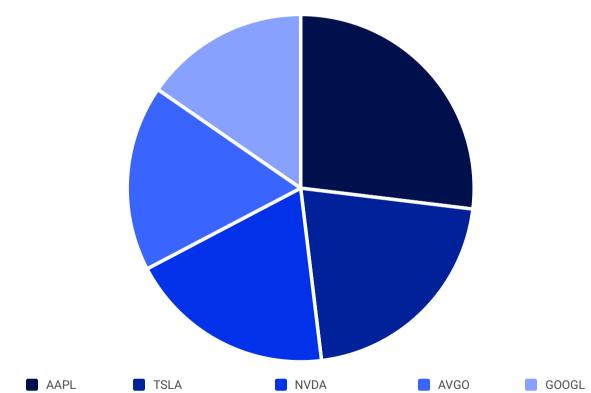
**Ending Value** 

**Total Return** 

**Sharpe Ratio** 

Max Drawdown Peak-to-trough decline during test period

Portfolio multiplier vs SPY's 2.46x benchmark Cumulative performance over 52-week period Risk-adjusted return efficiency metric



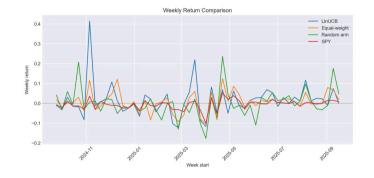
LinUCB demonstrated balanced exploration across all five assets, with AAPL receiving the highest allocation frequency.

## **Performance Visualization**



#### **Cumulative Value**

Portfolio growth trajectory compared against baseline strategies and SPY benchmark



#### **Return Resilience**

Weekly return distribution showing downside protection and volatility characteristics

Left panel shows cumulative value progression versus baselines | Right panel demonstrates weekly return resilience and risk profile

#### **Key Insight**

LinUCB achieved competitive returns with systematic risk management, demonstrating the viability of contextual bandits for tactical asset allocation in concentrated tech portfolios.

#### **Implementation Note**

The algorithm's balanced exploration-exploitation trade-off prevented over-concentration while capturing momentum signals across the five-asset universe.