



KaxaNuk
Sharing knowledge

KN Hack Kick-Off & Intro to Investment Research

Session 1

Content



- The Evolution of Investment Research
- Our Investment Research Process
- Common Strategies & Ideas
- Event Details

The Evolution of Investment Research

Observation → Fundamentals → Risk → Markets → Behavior →
Adaptation → Implementation

Markets as Signals — Dow, Bachelier & Nelson



Why they matter:

Treated markets as an information system before formal theory existed.

Key work:

- Dow (1889–1902) — Wall Street Journal Editorials
- Bachelier (1900) — *Théorie de la Spéculation*
- Nelson (1903) — *The ABC of Stock Speculation*

Core ideas:

- Prices embed collective information
- Trends reflect human behavior
- Uncertainty can be modeled probabilistically

“The market is a barometer of all conditions.”



If prices move first... what are
they reacting to?

Intrinsic Value — Graham, Dodd & Damodaran



Why they matter:

Transformed investing from price-watching into a research process based on value, discipline, and explicit assumptions.

Key work:

- Graham & Dodd (1934) — Security Analysis
- Damodaran (1994) — Investment Valuation
- Damodaran (2001) — The Dark Side of Valuation

Core ideas”

- Value exists independently of price
- Valuation is a model, not a number

“Valuation is a bridge between stories and numbers.”



If value exists... how do we
allocate capital across many
bets?

Risk Comes First - Markowitz



Why he matters:

Formalized portfolio construction as a mathematical research problem.

Key papers:

- Markowitz (1952) — Portfolio Selection
- Markowitz (1959) — Portfolio Selection: Efficient Diversification of Investments

Core ideas:

- Diversification is measurable
- Risk lives in covariance
- Portfolios > individual assets

“Investment research starts with how you allocate capital.”



If risk can be measured...
which risk deserves a
reward?

Pricing the Risk — Sharpe & Lintner



Why they matter:

Separated market compensation from research skill.

Key papers:

- Sharpe (1964) — Capital Asset Prices: A Theory of Market Equilibrium under Conditions of Risk
- Lintner (1965) — The Valuation of Risk Assets and the Selection of Risky Investments in Stock Portfolios and Capital Budgets

Core ideas:

- Beta earns a premium
- Performance attribution is born

“Not all risk is rewarded.”



If beta explains returns... what exactly are we being paid to research?

Measuring Skill — Jensen



Why they matter:

Jensen transformed “alpha” from intuition into a measurable residual.

Key papers:

- Jensen (1968) — The Performance of Mutual Funds in the Period 1945–1964

Core ideas:

- Alpha is performance after adjusting for risk
- **Most active managers fail to generate persistent alpha**
- Measurement precedes belief

“Investment research became a test, not a story.”



If alpha is rare and fragile,
should it be assumed—or
must it be proven?

Markets as a Discipline — Fama



Why he matters:

Forced investment research to start from skepticism, not belief.

Key papers:

- Fama (1970) — Efficient Capital Markets: A Review of Theory and Empirical Work

Core ideas:

- Prices reflect available information
- Alpha is rare
- Evidence beats intuition

“Alpha must be proven — not assumed.”



If markets are efficient... why
does evidence refuse to
behave?

Risk Is Multi-Dimensional — Ross



Why he matters:

Showed that returns must be driven by multiple independent sources of risk — even if we don't know what they are.

Key paper:

- Ross (1976) — The Arbitrage Theory of Capital Asset Pricing

Core ideas:

- No-arbitrage
- Multiple risk factors
- Theory before measurement

“Risk premia exist even when factors are unknown.”



If returns come from multiple risks... what are they, and how do we measure them?

Empirical Reality — Fama, French & Carhart



Why they matter:

Showed that returns have structure, even in efficient markets.

Key papers:

- Fama & French (1992) — The Cross-Section of Expected Stock Returns
- Fama & French (1993) — Common Risk Factors in the Returns on Stocks and Bonds
- Carhart (1997) — On Persistence in Mutual Fund Performance

Core ideas:

- Value, size, momentum
- Factor investing becomes scalable
- Research becomes systematic

“Markets are mostly efficient — but not simple.”



If factors persist... why are
they so hard to live with?

Human Behavior — Kahneman, Tversky & Shiller



Why they matter:

Explained why inefficiencies survive. Challenge to Rationality.

Key paper:

- Kahneman & Tversky (1979) — Prospect Theory: An Analysis of Decision under Risk
- Shiller (1981) — Do Stock Prices Move Too Much to Be Justified by Subsequent Changes in Dividends?

Core ideas:

- Loss aversion
- Cognitive bias
- Asymmetric risk preferences

“Risk is felt, not calculated.”



If mistakes are obvious... why
do they survive?

Limits to Arbitrage — Shleifer & Vishny



Why they matter:

Explain why being right isn't enough. Arbitrage is Risky and Costly.

Key papers:

- Shleifer & Vishny (1997) — The Limits of Arbitrage

Core ideas:

- Capital constraints
- Career risk
- Persistence of Anomalies

“Markets are stories with prices attached.”



So markets are rational... and
emotional. What framework
survives both?

Markets Adapt — Lo



Why he's the pivot:

Reframes markets as evolving systems, not static models. Reconciles the Efficient Market Hypothesis with Behavioral Economics.

Key papers:

- Lo (2004) — The Adaptive Markets Hypothesis: Market Efficiency from an Evolutionary Perspective

Core ideas:

- Strategies have life cycles
- Alpha is temporary
- Regimes matter

“Efficiency is not a state — it’s a process.”



If alpha decays... what does durable research look like?

Research Discipline in Practice — Asness



Why he matters:

Proved research survives markets through discipline.

Key papers:

- Asness (1997) — The Interaction of Value and Momentum Strategies
- Asness, Moskowitz & Pedersen (2013) — Value and Momentum Everywhere

Core ideas:

- Factors persist but cycle
- Robustness beats intuition
- Behavioral foundations matter

“The best strategies don’t look smart — they look resilient.”



If everything has exposure...
what is actually yours?

From Ideas to Portfolios — Paleologo



Why he matters:

Paleologo shows how investment research becomes real portfolios — by isolating residual alpha and enforcing risk discipline.

Key work:

- Paleologo (2021) — Advanced Portfolio Management: A Quant's Guide for Fundamental Investors
- Paleologo (2025) — The Elements of Quantitative Investing

Core ideas:

- Alpha must be residualized from known risk factors
- Portfolio construction is as important as idea generation

“Alpha is what remains after risk is removed.”



If clean signals are rare... who
finds them faster: humans or
machines?

Machine Learning — Halperin, Dixon, Bilokon



Why they matter:

They place machine learning inside the intellectual tradition of finance, not outside it — showing ML as a continuation of econometrics, time-series analysis, and stochastic control, rather than a black box replacement.

Key paper:

- Dixon, Halperin & Bilokon (2020) — Machine Learning in Finance: From Theory to Practice

Core ideas:

- Learning replaces assumptions about the data-generating process
- Finance becomes a design problem, not just a modeling problem

“Investment research becomes a system that learns.”



If learning systems can
model perception and
decision together, how
should we design investment
research going forward?

Quant 4.0 — Guo, Wang, Ni & Shum



Why they matter:

Research is no longer about finding isolated signals — it's about designing systems that continuously learn, adapt, and integrate human structure with machine scale.

Key paper:

- Guo, Wang, Ni & Shum (2022) — Quant 4.0: Engineering Quantitative Investment with Automated, Explainable and Knowledge-driven Artificial Intelligence

Core ideas:

- Alpha lives in representations, not raw signals
- Machines explore scale; humans impose structure
- Research is a system, not a model

“Modern investing is engineered, not discovered.”



Investment research did not
evolve by replacement.
It evolved by addition.



Every investment research
process is **discretionary** at
design and **systematic** at scale.

Quant thinking is what
connects the two.

Our Investment Research Process

Strategy Blueprint

The Synthesis of Evolution



How we think:

- **Financial Analysis** = The legacy of Graham & Damodaran.
- **Mathematical Models** = The logic of Markowitz & Sharpe.
- **Computational Algorithms** = The engineering of Halperin & Guo.

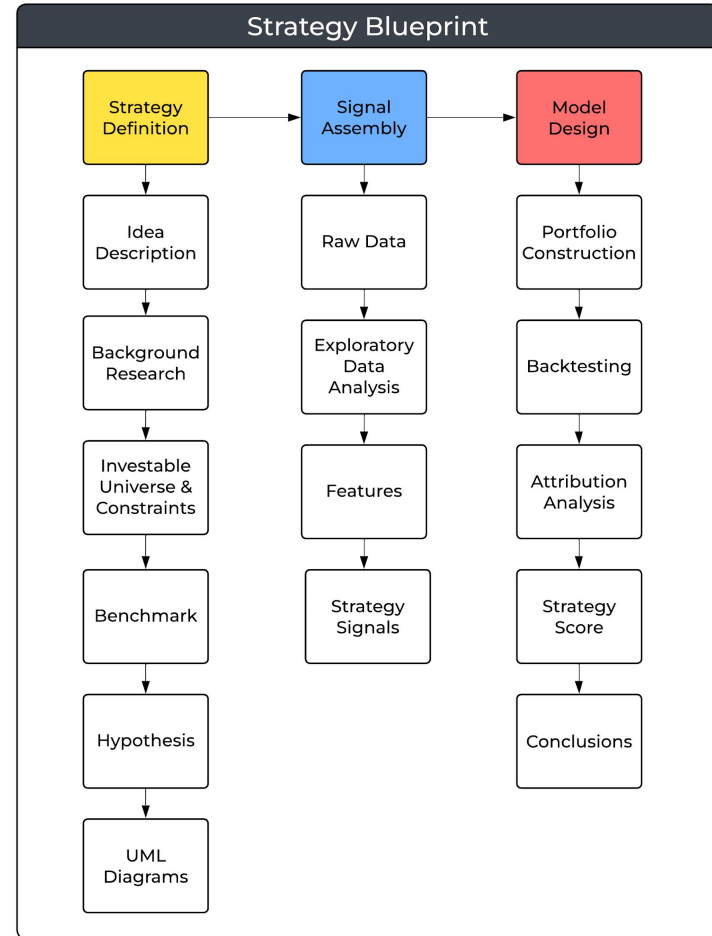
Strategy Blueprint

The Process Flow:

Strategy Definition: Moving beyond "intuition" to a formal hypothesis.

Signal Assembly: Transforming raw data into predictive signals.

Model Design: Implementing portfolio construction selection, timing and sizing.



Strategy Definition



Moving beyond "intuition" to a formal hypothesis.

Key Components

- **Investable Universe:** Where specifically are we looking for Alpha? (Equities, ETFs, or Crypto).
- **The Benchmark:** How do we prove we are better than a benchmark portfolio?
- **The Hypothesis:** We don't just trade; we test against specific market failures:
 - **Market Inefficiency:** Information lag.
 - **Risk Premium:** Rewards for liquidity or volatility.
 - **Behavioral:** Exploiting overreaction patterns.
 - **Valuation:** Finding value mismatch.

Signal Assembly



Transforming raw data into predictive signals.

Tools of the Trade

- **Data Curator:** Building structured databases from market and fundamental APIs.
- **Exploratory Data Analysis:** Identifying price trends and return distributions before modeling.
- **Feature Engineering:** Creating robust signals (Momentum, Volatility, Value) that survive out-of-sample testing.
- **The Rule:** "To **avoid overfitting**, we keep features constant until the conclusion of the research." Use your hypothesis idea to avoid data mining.

Model Design



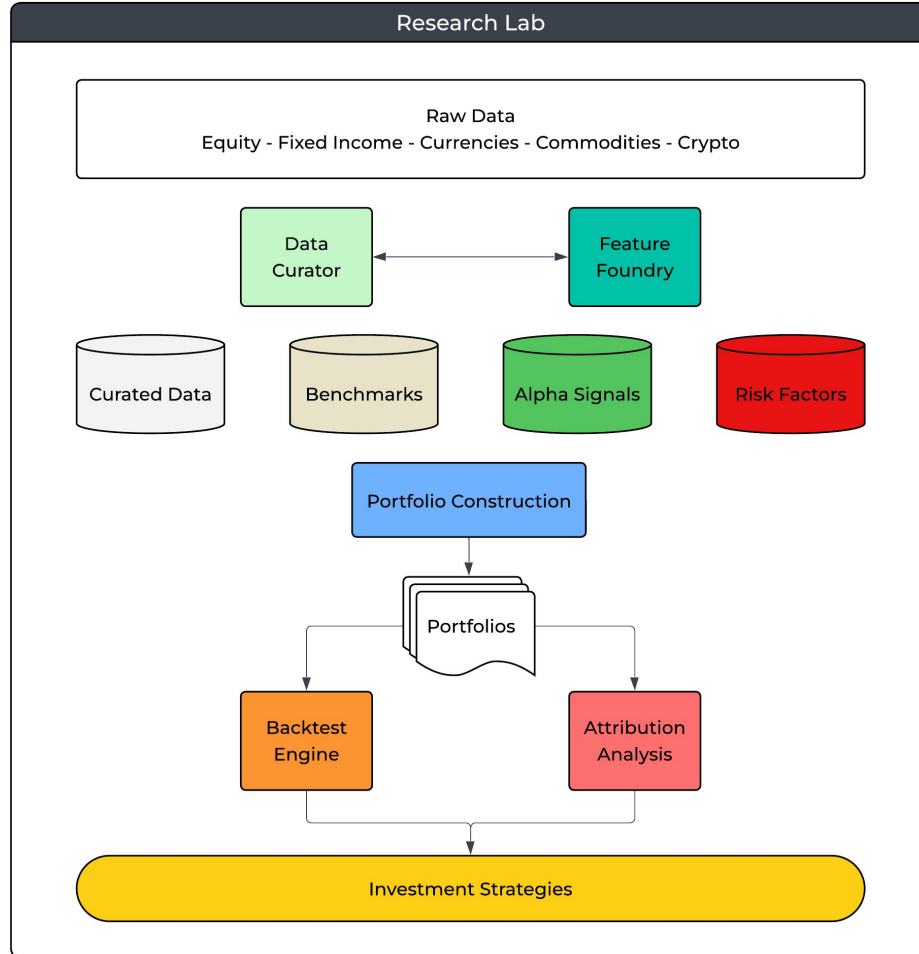
Implementing portfolio construction selection, timing and sizing.

Rigorous Implementation:

- **Constraints:** Respecting target holdings (e.g., 35) and weight limits (e.g., 20%).
- **Backtest Engine:** **Fully auditable Python-based process** accounting for slippage, fees, and commissions.
- **Attribution Analysis:** Decomposing returns to **prove the value added comes from Skill**, not just factor exposure.
- **The Final Output:** The **Strategy Score** — A quantitative verdict on implementability.

Research Lab

Uncover actionable ideas across asset classes, regions, and sectors.



Investment Strategy



KaxaNuk
Sharing knowledge

Backtest Dashboard

portfolio_weights_binary_filter

Analysis Period: 2010-03-10 to 2014-12-31

TOTAL RETURN

84.66%

ANNUALIZED RETURN

13.11%

SHARPE RATIO

0.943

SORTINO RATIO

1.307

ALPHA

4.225%

MAX DRAWDOWN

-19.36%

TOTAL COMMISSIONS

\$8,609

Portfolio Performance

Portfolio Performance Over Time



Portfolio Composition

Portfolio Composition - 2014-12-31 (Top 10)



Portfolio Weights Evolution

Commission Analysis

Returns Distribution

Drawdown Analysis

Annual Returns

Portfolio vs Benchmark Drawdown Analysis



Common Strategies Ideas

Brainstorming

Common Strategies and Investment Instruments

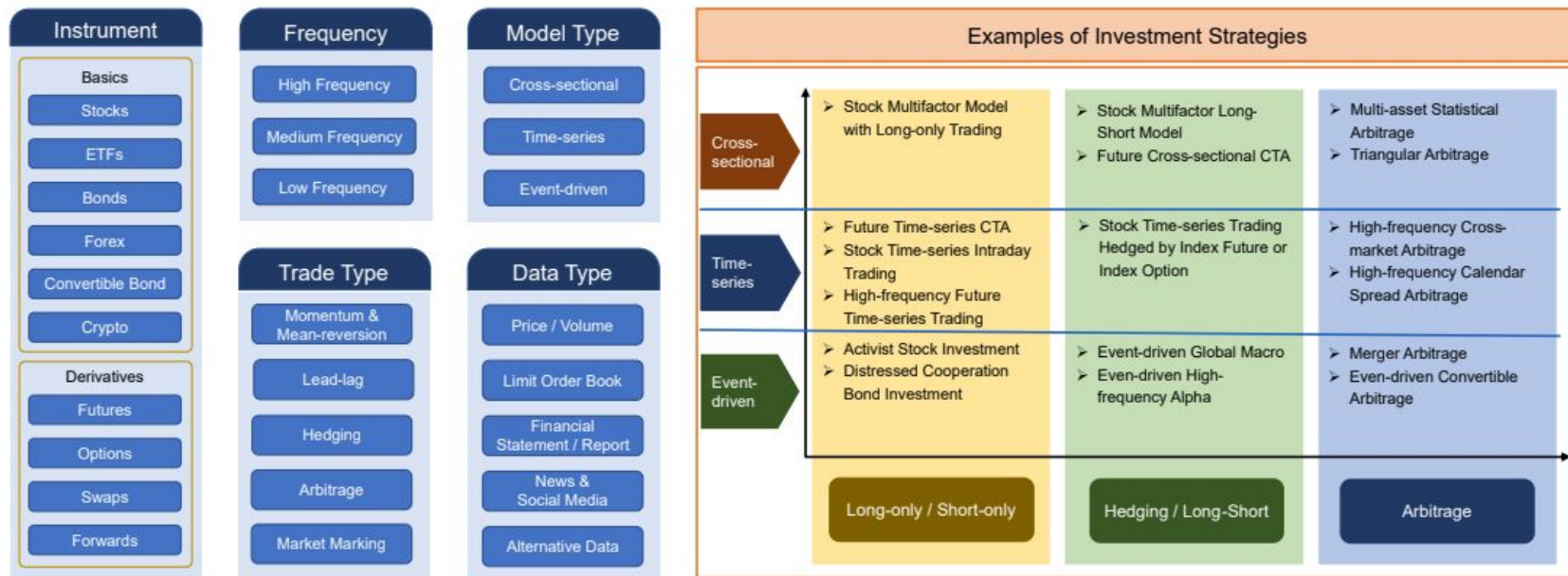


Figure 3: Classification of common strategies and investment instruments.

Popular Machine Learning Algorithms

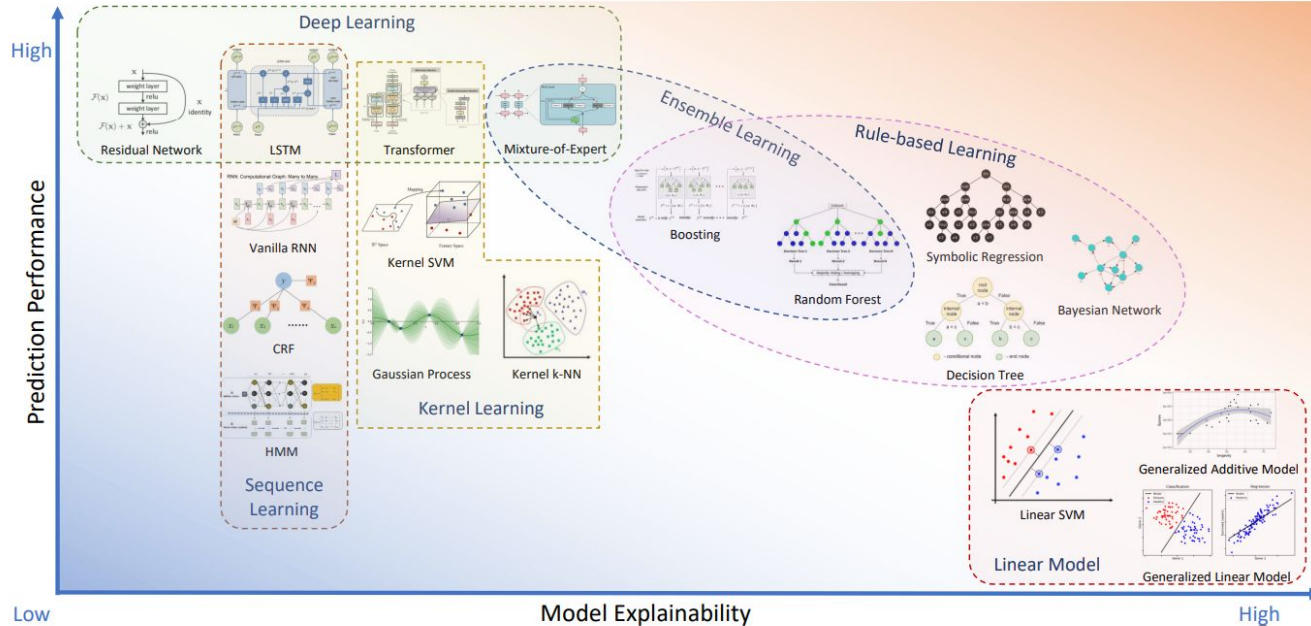


Figure 20: Comparison of popular machine learning algorithms according to prediction performance and model explainability. Part of this figure is cited from [130, 131, 132].

The Quality Value



- **Research Lineage:** Benjamin Graham + Aswath Damodaran.
- **The Concept:** Systematic screening for "Wonderful Companies at Fair Prices."
- **The Signal:**
 - **Value:** Low P/E or EV/EBITDA (The "Cheap" factor).
 - **Quality:** High Return on Equity (ROE) and Low Debt/Equity (The "Safety" factor).
- **The Hypothesis:** Markets overreact to bad news for good companies (Behavioral).
- **Data Needed:** Quarterly Fundamentals + Daily Closing Prices.
- **Frequency:** Monthly rebalancing.

Multi-Asset Trend Following



- **Research Lineage:** Charles Dow + Cliff Asness.
- **The Concept:** "The Trend is your friend." If an asset has been rising for 6–12 months, it is likely to continue.
- **The Signal:**
 - **Time-Series Momentum:** Comparing an asset to its own past (e.g., is Price > 200-day Moving Average?).
 - **Cross-Sectional Momentum:** Comparing assets to each other (e.g., buy the top 10% of ETFs by performance).
- **The Hypothesis:** Delayed reaction to information and "herding" behavior (Lo's Adaptive Markets).
- **Data Needed:** Daily/Weekly Closing prices for Stocks, Bonds, Commodities (ETFs).
- **Frequency:** Weekly or Monthly rebalancing.

Event-Driven: The Earnings Drift



- **Research Lineage:** Eugene Fama (Testing Efficiency).
- **The Concept:** Exploiting the Post-Earnings Announcement Drift (PEAD).
- **The Signal:**
 - Identify "**Earnings Surprises**".
 - Buy companies with a significant positive surprise.
- **The Hypothesis:** Market prices are efficient but not instant. It takes weeks for investors to fully process a fundamental change in a company's trajectory.
- **Data Needed:** Earnings dates, fundamentals, and Daily Prices.
- **Frequency:** Mid-frequency (Trades held for 10–60 days).

The Low Volatility Anomaly



- **Research Lineage:** Harry Markowitz + William Sharpe.
- **The Concept:** CAPM says "higher risk = higher return," but historical data often shows that Low Volatility stocks outperform high-risk stocks over long periods.
- **The Signal:**
 - Identify stocks with the **lowest 36-month realized volatility or Beta**.
 - Invert the weights (Risk Parity).
- **The Hypothesis:** Leverage constraints and lottery-preference bias (Investors overpay for "exciting" stocks and ignore "boring" ones).
- **Data Needed:** Daily Closing prices (to calculate variance/covariance).
- **Frequency:** Quarterly or Semi-annual rebalancing.

Event Details

The Challenge

Are You Ready to Test Your Skills?



We are excited to announce the **KN Hack Research Challenge**, an opportunity for aspiring quants, data scientists, and finance enthusiasts to showcase their analytical prowess and innovative thinking!

Whether you're a student, a professional, or simply passionate about quantitative finance, this challenge is for you!

Challenge Overview



Participants will be tasked with solving a real-world **investment strategy** applying rigorous scientific methods to investing, combining **financial analysis**, **mathematical models**, and **computational algorithms**.

Key Details



Training Sessions: Monthly online session starting on January 2026 until May

Registration Deadline: May 31st 2026

Event: June 11th to 13th 2026

Location: Puebla, Mexico

Eligibility: Open to teams of 2 to 4 members

Judging Criteria



Innovation: Originality and creativity in approach

Technical Rigor: Use of appropriate quantitative methods

Clarity: Quality of presentation and explanation of findings

Practicality: Applicability of the solution to real-world scenarios

Performance: Risk adjusted returns and consistency

Overfitting: Data segmentation, walk-forward analysis,
cross-validation

Why Participate?



Enhance Your Skills: Gain hands-on experience in quantitative analysis and problem-solving.

Network with Professionals: Connect with industry experts and participants.

Win Prizes: Compete for exciting rewards that could kickstart your career in finance.

3 Topics



Equities: Stock Picking

- Generate alpha through high-conviction stock selection.
- Deploy quantitative, technical, or fundamental strategies.
- Outperform the benchmark within a designated investment universe.



Multi-Asset: ETF Allocation

- Construct robust portfolios across global equities, bonds, real estate, and commodities.
- Test skills in dynamic asset allocation and risk management.
- Design strategies that optimize returns beyond traditional models.



Crypto: Digital Assets

- Navigate the volatility of the digital asset economy.
- Apply strategies to top-tier cryptocurrencies.
- Capture growth across the broader crypto ecosystem.



Disclaimers



The content of this document is strictly informative and does not constitute an offer or recommendation of KaxaNuk S.C. to buy, sell or subscribe any kind of securities, or to perform specific transactions. KaxaNuk S.C. is not responsible for the interpretation given to the information and /or content of this document. KaxaNuk S.C. does not accept or will accept any liability for losses or damages resulting from investment decisions that would have been based on this document. The persons responsible for the preparation of this content certify that the opinions stated reflect their own point of view and do not represent the view of KaxaNuk S.C. nor of its officials. This document is based on publicly available information which are considered reliable, however KaxaNuk S.C. makes no warranty regarding its accuracy or completeness.

Contact
research@kaxanuk.mx

