MULTIVERSE

Quantum Computing for Finance



Summary

- 1 Introduction Multiverse Computing
- Multiverse benchmarks and use-cases for Finance
- 3 Singularity Portfolio optimization
- Singularity Python Frontend demo



MULTIVERSE - Global Quantum Software Revolution Winner

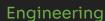
• Top-3 Global Player **Largest & Fastest-Growing** ~\$8B Medium-Term Market Opportunity << >> • ~\$45B Long-Term Opportunity in Quantum Software **Quantum Software Company Globally** · Offices in Spain, Canada, France & Germany More Blue-Chip Clients & Real-World << BANK OF CANADA

BANOUE DU CANADA 🔲 - BASF BOSCH REPSOL **Revenue Generating Use Cases than Anyone High-Growth Recurring** €6.7M Total Contract Value Sold in 2021 << >> · Significant Projected Revenues in 2022 and Beyond **Revenue Profile** Revolutionary Hardware-Agnostic SaaS Platform **Highly-Scalable Hardware Agnostic** << · 22 Patents Filed in 2021 **Singularity SaaS Platform** • 24 More Patents Planned in 2022 **Visionary Industry-Leading World-Class** 9,000+ 30% 25% << **Team** Citations PhDs Women **Nationalities** BCG Gartner. **₹** Sifted << **Global Industry Accolades** >> "Cool Vendor" 2022 100x Speed Improvements in Quantum Startup for Banking 2021 Financial Services 2020 **Clean Energy Green Al Fundamental Technology Platform and** 44 **Unrivalled Expertise to Accelerate ESG** Health & Life Industrial **Materials Strategies Climate Technology Operations Sciences** QUANTO NATION /inveready **Early Specialist Expert Investors** << >> **JME VENTURES**

We Deliver Value to Blue-Chip Customers Across an Unparalleled Set of **Industry Verticals**

Aerospace Predictive Maintenance





Optimize Component **Functionality**



Forex Trading



Manufacturing / Supply Chain

Quantum Digital Twin



Price Derivatives Accurately

Finance



Renewable Energy

Simulate Energy Demand



Life / Healthcare

Predict Patient Health Crises



Market Simulation



Logistics

Routing Optimization





Government

Green Hydrogen











































Customers and Qualified Prospects





Strictly Private and Confidential

Benchmarks and use-cases for Finance

Singularity Toolbox



APPS

Integrate our cutting edge solutions straight out of the box



SINGULARITY FINANCE

- Asset management
- Derivatives pricing
- Risk management



Develop your own quantum applications built on top of Singularity core - No quantum experts required!

CORE



SINGULARITY MACHINE LEARNING

- Leverage a suite of powerful and easy to use quantum machine learning algorithms
- Classification, regression, deep learning



SINGULARITY OPTIMIZATION

- Quickly build and solve optimization problems
- Solve challenging discrete optimization

Harness the power of quantum technologies backends





TENSOR NETWORK (Quantum Inspired)

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QUANTUM

- Run on the latest quantum hardware from our partners
- Leverage true quantum power

Summary of Application Areas

Static Portfolio Optimization



Dynamic Portfolio Optimization



Asset Management

Index Tracking

ally

Algorithmic FX Trading

New Application Areas

□ • BASF
We create chemistry

Collateral Optimization Optimal Trade Execution

Derivatives Pricing

High Dimensional Basket Options Pricing (Heston)



Bermudan Swaptions
Pricing

Large European Bank **Autocallable Notes**

Large South
American Bank

Risk Management and Macroeconomics

Credit Downgrade
Prediction



Fraud Detection

Data from Citi

Cryptocurrency Adoption Networks



Financial Crash Prediction





Derivatives Pricing

Experts in quantum inspired deep learning approaches



Derivatives Pricing at a Glance



Tensor Network

- Real advantages and value **TODAY**
 - Efficient Less parameters -> Faster learning
 - Reliable More precise pricing and tighter confidence
 - Scalable Handle models past the breaking point of current standard pricers.
- Deep learning with Tensorized Neural Networks (TNN)

What can we price:

- High dimensional models
- American style payoffs: New accurate approach
- Path dependent payoffs



Quantum

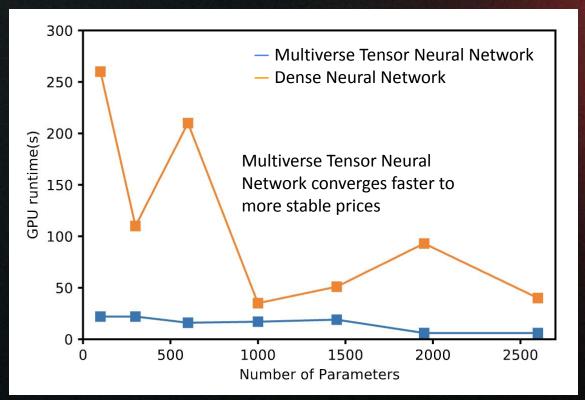
- Quantum Monte Carlo (QMC)
- Prepare for the quantum winter
- Build hardware ready infrastructure
- Evaluate performance of current hardware
- State-of-the-art circuits for path dependent options





High Dimensional Baskets - Heston Model

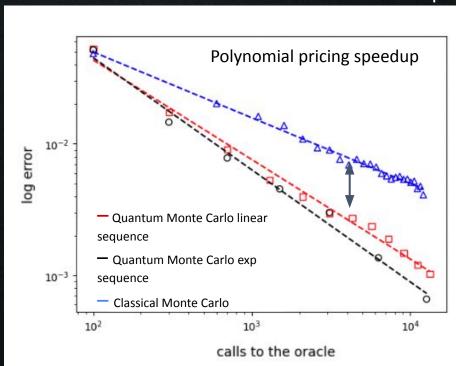
- Less parameters with better accuracy than current state-of-the-art methods
 - 1950 classical parameters, 600 tensor network parameters
- Faster and more stable prices compared to state-of-the-art methods

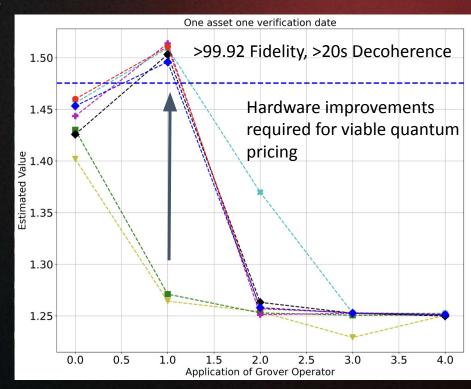




Autocallables - Quantum Monte Carlo

- Study potential of applying quantum monte carlo for path dependant barriers
- Novel circuit to encode Autocallable note payoffs





In progress

- Study determining hardware conditions for accurate pricing on real hardware



Risk Management

Experts in quantum machine learning for unbalanced data



Risk Management at a Glance

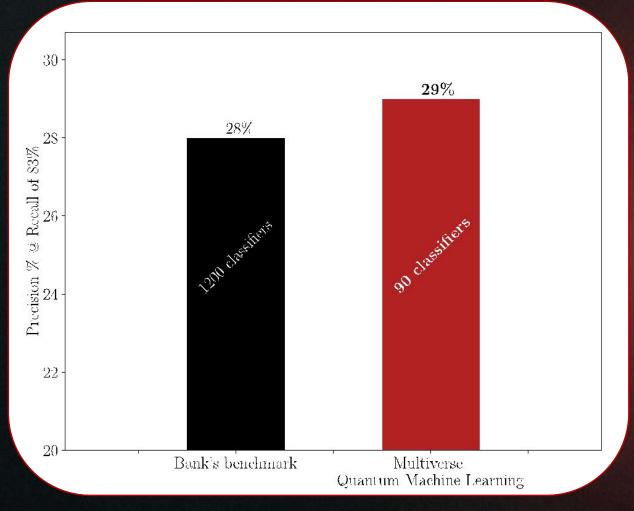
Using state-of-the-art quantum machine learning we **improve the accuracy and efficiency**.

- Fallen Angels Credit downgrades
 - Improved prediction accuracy with x13 smaller model
- Credit Card Fraud
 - 100x faster, improved accuracy saves up to \$80M/year
- Predicting Financial Crashes
 - Perturbation analysis of financial network stability
 - Analyse large networks classically intractable
- Cryptocurrency Network Adoption
 - Analyse complex network formation dynamics
 - Analyse large networks classically intractable





Fallen Angels: Multiverse predicts credit scoring better



- Multiverse Quantum
 Machine Learning
 catches the same
 number of credit
 downgrades* with
 more precision and less
 false positives
- Multiverse solution is way <u>more interpretable</u> with much less classifiers used to predict the credit downgrade



*these are called Fallen Angels, for example clients/bonds/loans whose solvency is at risk.

Asset Management

Leveraging current quantum technologies to deliver real value for complex optimizations







Asset Management at a Glance

Using quantum and quantum inspired technologies we can solve the most challenging optimization problems and machine learning tasks.

Algorithmic Trading

+38% increase in returns and reduced risk



Index tracking

- x2 better risk return and 25% less assets
- Apply to ETF construction



Portfolio Optimization

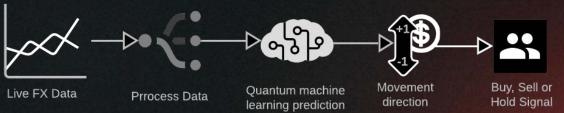
- Multiverses' hybrid quantum solver provides superior results
 - Optimal solutions, reduced computational overhead
 - Exactly solve for discrete asset allocations
- Dynamic optimization solves computational intractable problems

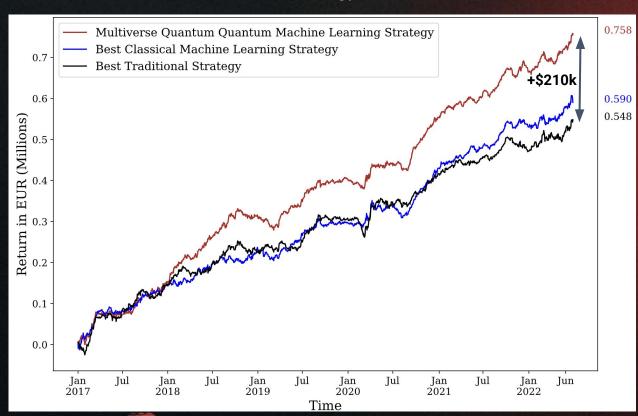






Algorithmic Trading - More profitable trades





Why Quantum?

- +38% increase in returns compared to client trading strategy
- +28% increase compared to classical machine learning
- +\$1m for client trading (\$5million notional)
- Return increases are consistent and sustained year-on-year
- More reliable signals -> lower risk

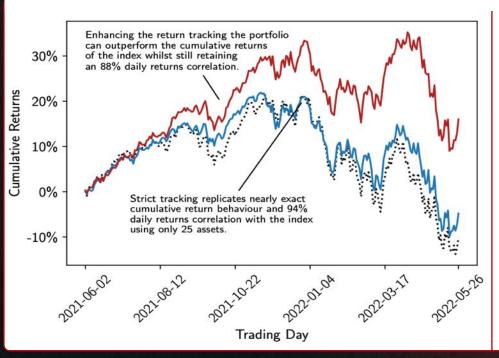
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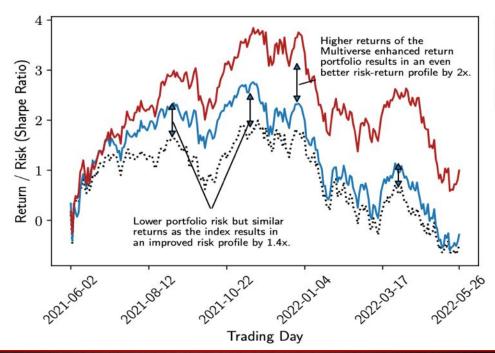
*Multiverse enhanced classical machine learning toolkit.



Index Tracking with smaller portfolios





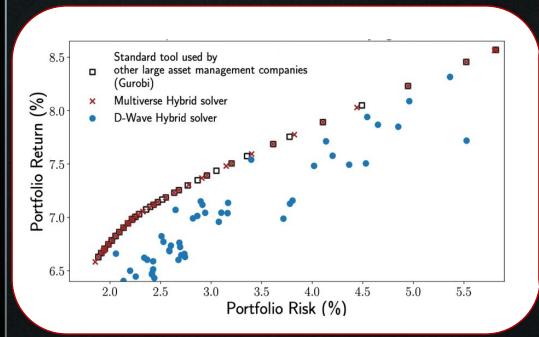


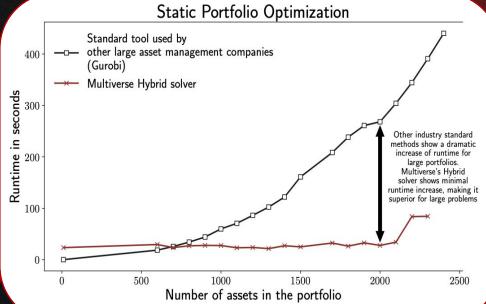
Why Multiverse Quantum?

- Achieve high degree of tracking (94%) with x4 less assets
- Cardinality constrained optimization challenging for classical solvers easy for quantum
- Up to 2x reduction in risk relative to full index



Multiverse Hybrid Optimization





Why Multiverse Quantum?

- Same high quality solutions compared to industry standard solvers (Gurobi)
- Outperforms other best-in-class quantum solvers (DWave Leap Hybrid)
- Scalable to large numbers of asset 500+
- Faster for large portfolios Minimal computational overhead



*these benchmarks are only applied to <u>static</u> portfolio optimization, given that the classical **state-of-the-art solutions used in the** market cannot find any solution in more complex <u>dynamic</u> portfolio optimization problems, where Multiverse Hybrid solver does find optimal solutions.

New Directions and Exploration

Pioneer and develop new quantum applications and innovations



New Directions and Exploration - A Taster

Portfolio Optimisation

Derivatives Pricing

Algorithmic Trading

Optimal Trade Execution

- Dynamic optimization explodes over many timesteps for classical
- Market impact of trades
- Reduce slippage

Collateral Optimization

- Optimize asset collateral
- Taking into account problem explodes for classical solver
- Quantum give faster and optimal solutions

ESG Optimization

- Include new ESG factor constraints and other complex investor preferences
- Solve for classically challenging constraints and large portfolios

Parametric Pricers

- Holy grail of derivatives pricing.
- Model learns the full pricing equation
- Incredibly fast pricing

Gaussian Processes and Quantum Kernels for Pricing

- Hot topic in finance
- Gaussian process have interpretable uncertainty
- More accurate prices and confidence

Quantum Reinforcement Learning for Pricing/Hedging

- Another hot topic in finance
- Price and hedge using optimal behaviour

Quantum Reinforcement Learning

- Learn dynamic trading behaviours
- Quantum could compactly represent decisions and actions
- More profitable trades

Statistical Arbitrage

- Quantum methods natural capture correlation
- Leverage to find arbitrage strategies

Our Finance Publications

- 1. Quantum-Inspired Tensor Neural Networks for Option Pricing https://arxiv.org/abs/2212.14076
- 2. Financial Index Tracking via Quantum Computing with Cardinality Constraints https://arxiv.org/abs/2208.11380
- 3. Quantum Portfolio Optimization with Investment Bands and Target Volatility https://arxiv.org/abs/2106.06735
- 4. Use Cases of Quantum Optimization for Finance https://arxiv.org/abs/2010.01312
- 5. Dynamic Portfolio Optimization with Real Datasets Using Quantum Processors and Quantum-Inspired Tensor Networks https://arxiv.org/abs/2007.00017
- 6. Forecasting financial crashes with quantum computing https://arxiv.org/abs/1810.07690
- 7. Hybrid Quantum Investment Optimization with Minimal Holding Period https://arxiv.org/pdf/2012.01091.pdf

- 8. Financial Risk Management on a Neutral Atom Quantum Processor https://arxiv.org/abs/2212.03223
- 9. Towards Pricing Financial Derivatives with an IBM Quantum Computer https://arxiv.org/abs/1904.05803
- 10. Towards Prediction of Financial Crashes with a D-Wave Quantum Computer https://arxiv.org/abs/1904.05808
- 11. Quantum computing for finance: overview and prospects https://arxiv.org/abs/1807.03890
- 12.Quantum portfolio value forecasting https://arxiv.org/abs/2111.14970
- 13. Quantum-Inspired Tensor Neural Networks for Partial Differential Equations https://arxiv.org/abs/2208.02235
- 14. Variational Quantum Continuous Optimization: a Cornerstone of Quantum Mathematical Analysis https://arxiv.org/abs/2210.03136
- 15. The Future of Quantum Finance https://www.proquest.com/openview/3f8a699a0ed484833e9d7e6d3b34c98c/1?pq-origsite=gscholar&cbl=48426





Singularity Toolbox



\PPS

Integrate our cutting edge solutions straight out of the box



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Singularity Portfolio Optimization Excel Plugin v1.2

- Installs directly into Excel for Windows, super easy and straightforward!
- Target user: Portfolio Manager

Inputs

- Forecasted returns
- Covariances
- Resolution: Precision of output holdings
- Risk aversion
 - 0 value: Aim for best return, no matter the risk
 - +infty value: Diversify portfolio across unrelated assets, at the expense of big reward
- Investment bands

Outputs

- Optimal Holdings
- Key Performance Indicators (KPI):
 - Return: return of the portfolio
 - Volatility: Risk of the portfolio
 - Sharpe ratio: Return per unit of risk



Solvers







Multiverse Hybrid

- Quantum + Classical
 - Solve discretized convex problems
 - Fast

D-WAVE Leap Hybrid

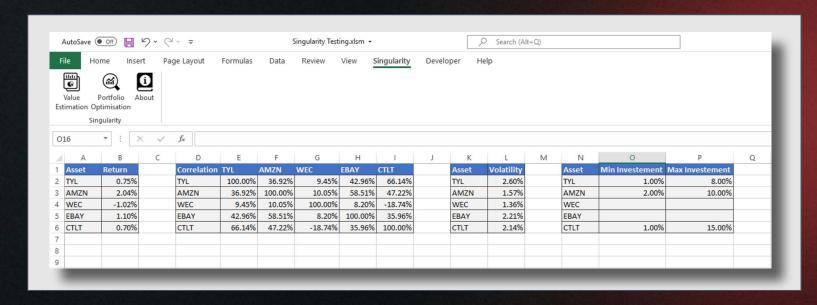
- Hybrid solver
 - Solve any QUBO
 - Slow

Classical solver

- No quantum
 - Solve discretized convex problems
 - Fast



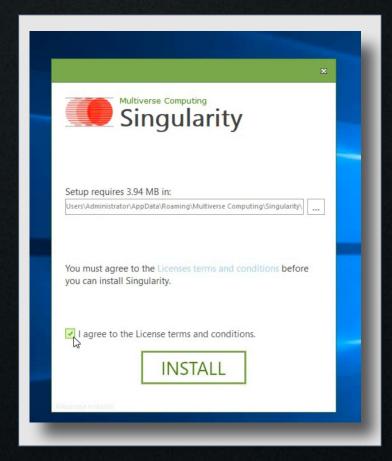
Features

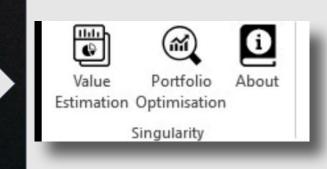


- Finds optimal Portfolio for a target risk
- User interacts through Excel interface
- Inputs: Returns, Volatilities, Correlations, investment bands, total investment, default max investment per asset, target volatility
- Outputs: holdings, sharpe ratio, energy, total return, volatility



Installer





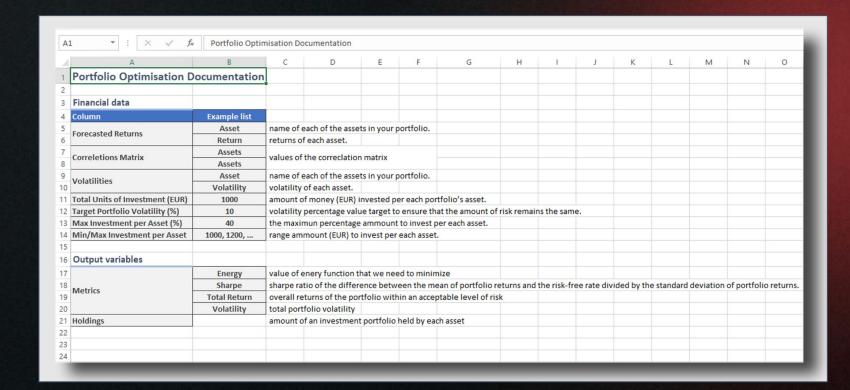


login





Documentation





Submit optimisation job

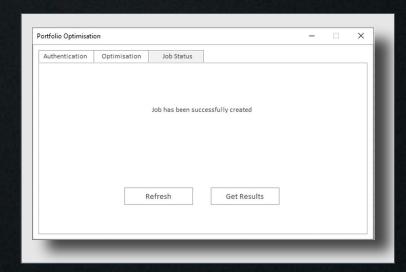
| A | В | C | D | E | F | G | Н | 1 | J | K | L | M | N | 0 | Р |
|-------|--------|---|-------------|---------|---------|---------|---------|---------|---|-------|------------|---|-------|-----------------|-----------------|
| Asset | Return | | Correlation | TYL | AMZN | WEC | EBAY | CTLT | | Asset | Volatility | | Asset | Min Investement | Max Investement |
| TYL | 0.75% | | TYL | 100.00% | 36.92% | 9.45% | 42.96% | 66.14% | | TYL | 2.60% | | TYL | 1.00% | 8.009 |
| AMZN | 2.04% | | AMZN | 36.92% | 100.00% | 10.05% | 58.51% | 47.22% | | AMZN | 1.57% | | AMZN | 2.00% | 10.009 |
| WEC | -1.02% | | WEC | 9.45% | 10.05% | 100.00% | 8.20% | -18.74% | | WEC | 1.36% | | WEC | | |
| EBAY | 1.10% | | EBAY | 42.96% | 58.51% | 8.20% | 100.00% | 35.96% | | EBAY | 2.21% | | EBAY | | |
| CTLT | 0.70% | | CTLT | 66.14% | 47.22% | -18.74% | 35.96% | 100.00% | | CTLT | 2.14% | | CTLT | 1.00% | 15.009 |



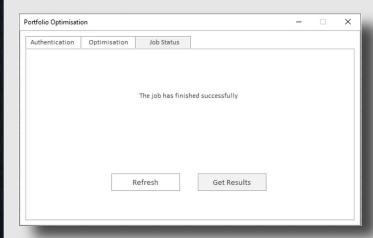
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| | Job Status | | |
|-------------------|-------------------|--|--|
| nisation Data'! | \$A\$2:\$B\$6 - | Total Units of Investment | 1000 |
| ix nisation Data' | \$D\$1:\$I\$6 - | Target Portfolio Volatility (%) | 10 |
| nisation Data' | \$K\$2:\$L\$6 - | Max Investment per Asset (%) | 30 |
| | | Min/Max Investment per Asset | \$N\$2:\$P\$6 - |
| | Opt | imise | |
| | ix nisation Data' | ix nisation Data'!\$D\$1:\$I\$6 - nisation Data'!\$K\$2:\$L\$6 - | ix nisation Data (\$\mathcal{9}\mathcal{9}\mathcal{1}\); nisation Data' (\$\mathcal{1}\); nisation Data' (\$\mathcal{1}\); N\$ and note that the second of |

Job submission & Results sheet









| Res | Results | | | | | | | | | | | |
|-----|---------|---------------------|-------------|---|----------|------|------|-------|-------|------|---|--|
| | 4 | Α | В | C | D | E | F | G | Н | 1 | J | |
| | 1 | Metrics | | | Holdings | | | | | | | |
| | 2 | energy | 0.002129214 | | | TYL | AMZN | WEC | EBAY | CTLT | | |
| | 3 | Sharpe | 0.500277235 | | 58:08.4 | 0.08 | 0.1 | 0.159 | 0.511 | 0.15 | | |
| | 4 | Total Return | 1.007676994 | | | | | | | | | |
| | 5 | Volatility | 0.015345479 | | | | | | | | | |
| | 6 | | | | | | | | | | | |
| | 7 | | | | | | | | | | | |



Now let try to do a portfolio optimization on real quantum computer, but with the Python Frontend!

Singularity Portfolio Optimization Demo

In this notebook we demonstrate how to use the Singularity Portfolio Optimization to solve a common portfolio problem.

Data specification

The examples covered in this section make use of real market data for three of the largest market-cap stocks in the S&P 500 index as of January 2023, selected purely for demonstration. These are Exxon Mobil Corporation (XOM), Johnson & Johnson (JNJ), and Apple Inc. (AAPL). The publicly available input data included below reflects the annual return for these assets across 2022 and the covariance over that period based on daily price changes.

```
In [1]: import logging
  import singularity.portfolio_optimization as po
  from dataclasses import asdict
  from plot_bar_chart import plot_bar
```

Add the logger.

In [2]: po.add_logger(logging.INFO)

First, specify the assets names and store them in the Assets object.

In [3]: asset_names = ["XOM", "JNJ", "AAPL"]
assets = po.Assets(asset_names)

Then, specify the expected return of each asset.

