

# Leveraging Julia for Advanced Data Visualization in Pharmaceutical Research

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# Agenda

1. The Importance of Data Visualization
2. Overview of the Current Visualization Landscape
3. Introduction to Julia
4. Visualization Tools within the Julia Ecosystem
5. Comparison of Performance Metrics
6. Regulatory Considerations
7. Where Julia fits today ?
8. Future Potential



# Why Data Visualization Matters in Pharma

1

## Complex clinical data

Clinical trials generate high-volume, high-dimensional data that requires advanced visualization to uncover patterns, trends, and anomalies efficiently.

2

## Regulatory transparency

Clear and traceable visualizations enhance data interpretability, supporting regulatory review, audit readiness, and decision confidence.

3

## Faster insights & decision-making

Interactive and responsive visualizations enable quicker insight generation, allowing teams to make timely, data-driven decisions.

# Current Visualization Landscape



## SAS, R, Python dominate

SAS, R, and Python form the backbone of current pharmaceutical analytics, covering regulated reporting, statistics, and exploratory visualization.



## Performance & scalability challenges

As data volumes grow, performance limitations and scalability challenges become more visible in complex and interactive visualization workflows.



## Growing data complexity

Modern clinical and operational datasets are increasingly high-dimensional, longitudinal, and heterogeneous, requiring more advanced visualization approaches

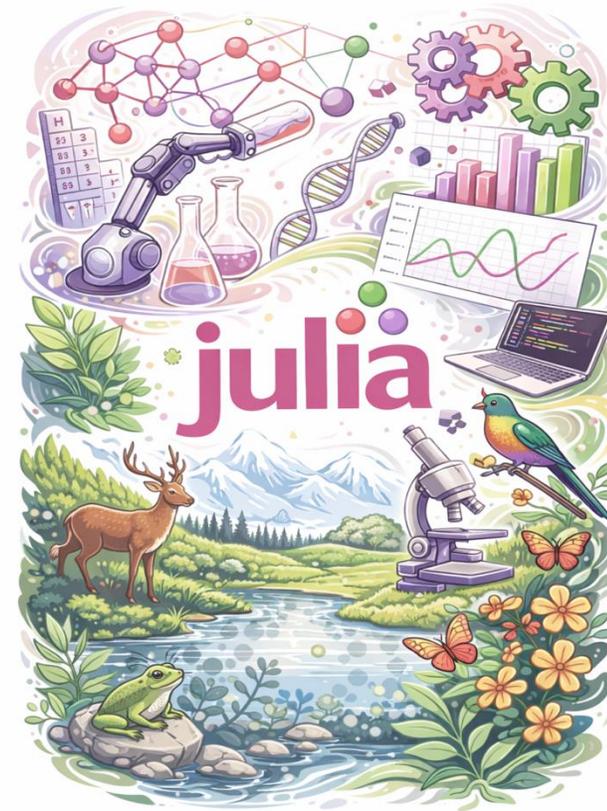
# Introducing Julia



- High-performance language
- Designed for numerical & scientific computing
- Open-source

# Why Julia Now?

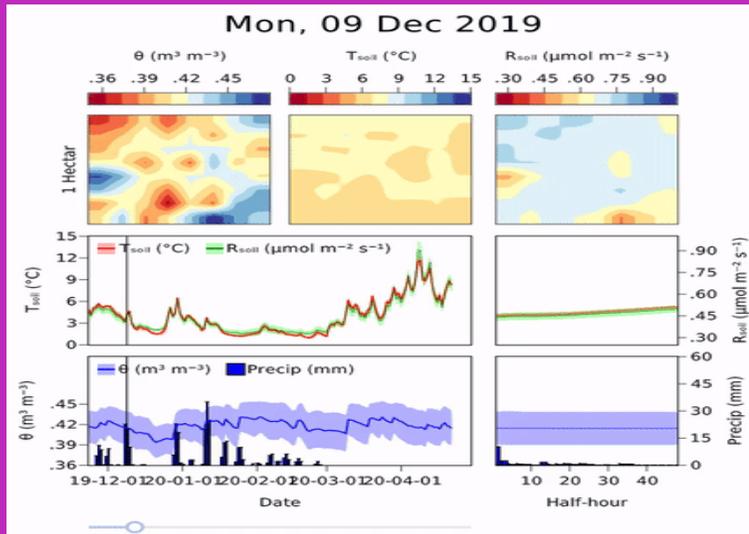
- Speed comparable to C
- Python-like simplicity
- R-like statistical strength



# Julia Ecosystem for Visualization

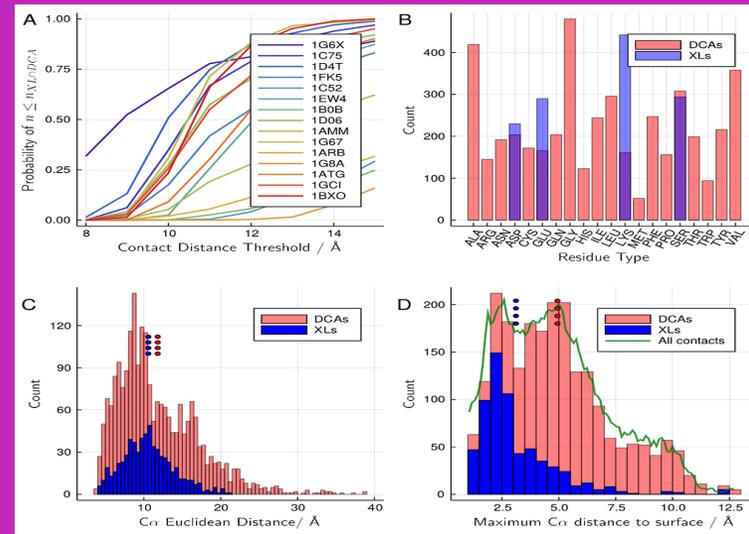
## 1 Makie.jl

Makie.jl enables high-performance, interactive, and GPU-accelerated visualizations for large and complex datasets.



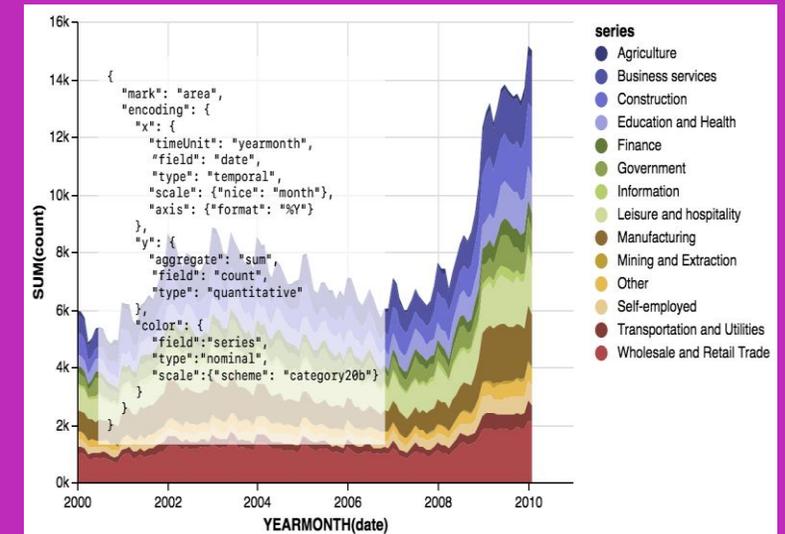
## 2 Plots.jl

Plots.jl provides a simple and flexible plotting interface that supports multiple backends for quick and consistent visualizations.



## 3 VegaLite.jl

VegaLite.jl enables declarative, grammar-of-graphics visualizations suitable for dashboards and exploratory analytics.

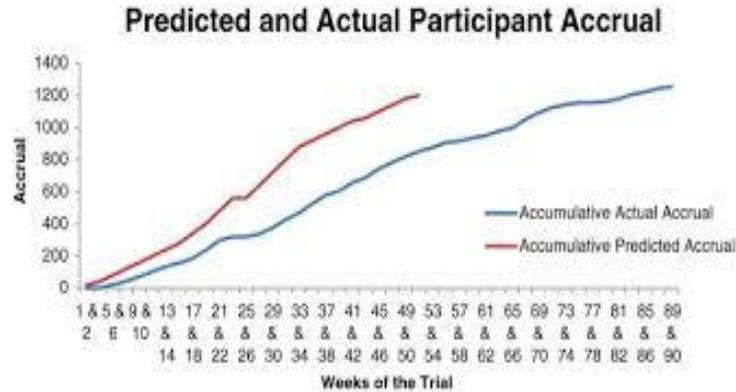


# Clinical Trial Data Visualization

( from raw data to actionable insights )

- Patient enrolment trends

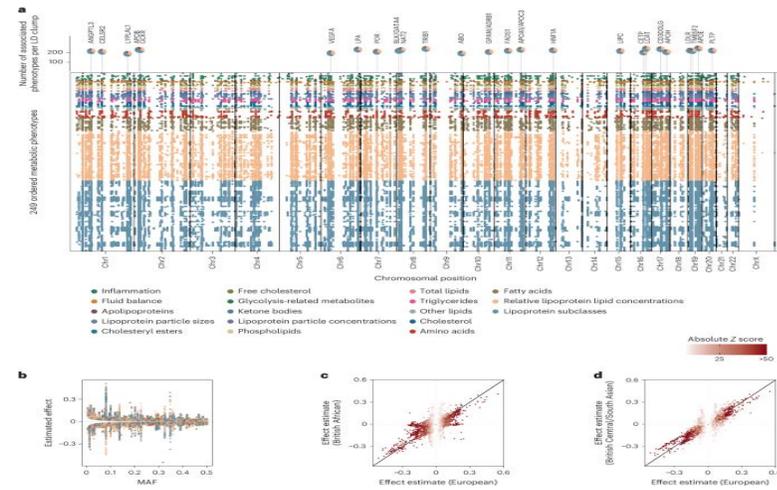
- Early identification of slow-enrolling sites



Cumulative enrollment over time (line chart)

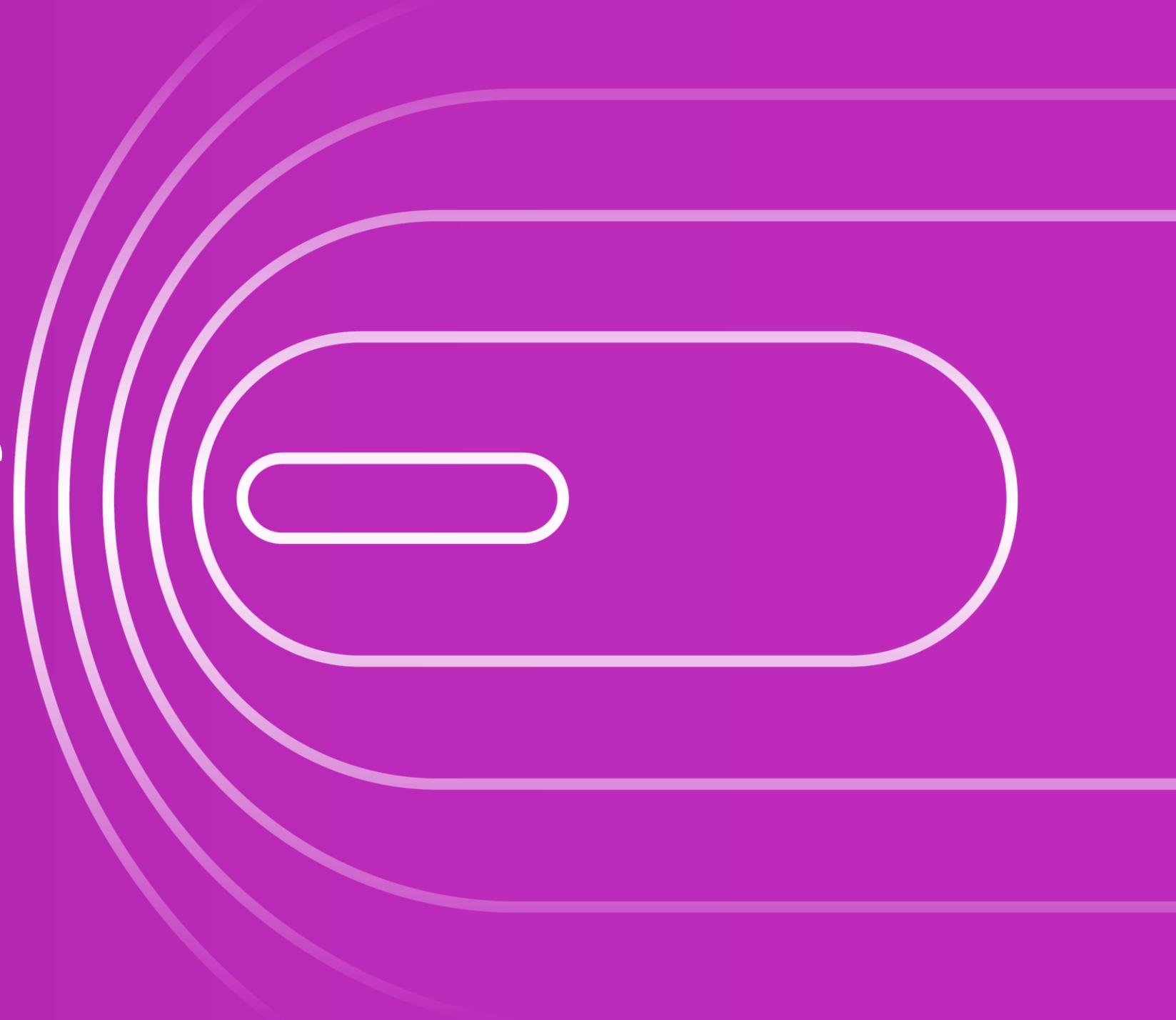
- AE distributions

- Rapid drill-down into high-risk events



AE trends over time or by treatment arm

# Performance Comparison



# Julia Compared to R and Python: Performance & Interactivity

## *Julia vs R vs Python*

- ✓ Julia delivers near C-level performance with high-level syntax
- ✓ R and Python often rely on optimized external libraries
- ✓ Julia avoids frequent language switching for performance tuning

**Ex:**

“In Julia, the same code used for prototyping can often be used for high-performance execution, without rewriting in C++.”

## *Large datasets*

- ✓ Designed for efficient memory usage and parallel execution
- ✓ Handles millions of rows smoothly for analytics and visualization
- ✓ Performance scales as data size increases

**Ex:**

“Visualizing longitudinal lab data across thousands of subjects and visits without sampling or downscaling.”

## *Real-time interactivity*

- ✓ Responsive zoom, pan, and filtering on large datasets
- ✓ Interactive dashboards update without long refresh delays
- ✓ Suitable for exploratory and operational monitoring

**Ex:**

“Real-time exploration of enrollment trends or AE distributions during ongoing trials.”

# Pseudo Code Example

Load data

Transform

Visualize



# Regulatory Considerations

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GxP awareness

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Validation strategy

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Exploratory vs submission

# Where Julia Fits Today

- **Exploratory analytics**
- **Advanced visualization**
- **POCs**



## Industry Use Case:

- High-Performance Pharmacology Simulations
- Advanced Disease Modeling and Predictive Analytics

# Future Potential

Hybrid  
workflows

Integration  
with Python/R

High-performance  
analytics

1. Start small
2. POCs
3. Training & governance

# Adoption Strategy

# Key Takeaways

**Julia is  
promising**

**Not replacement  
but complement**

**Future-ready**

Q & A ?

Thank you!

# Reference

1. <https://juliapackages.com/>

2. <https://github.com/>