

# GridAI: Cloud-Based Machine/Deep Learning For Power Grid Data Analytics

sdmay21-23

Faculty Advisor & Client:

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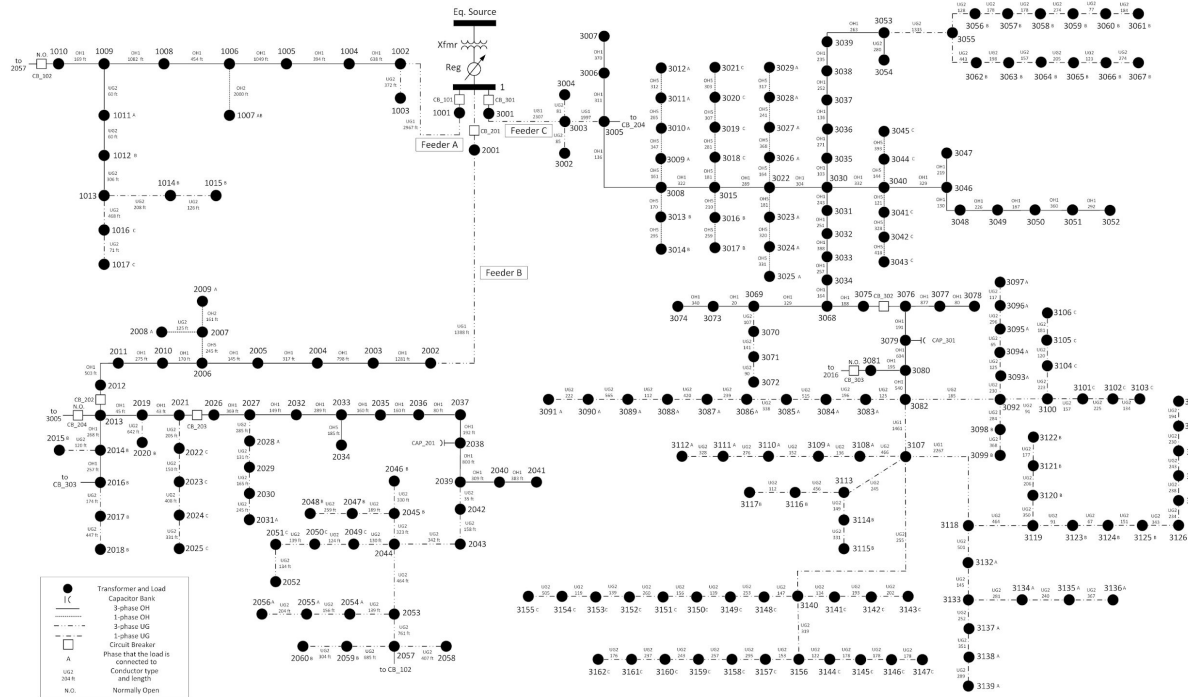
Justin Merkel

Patrick Wenzel

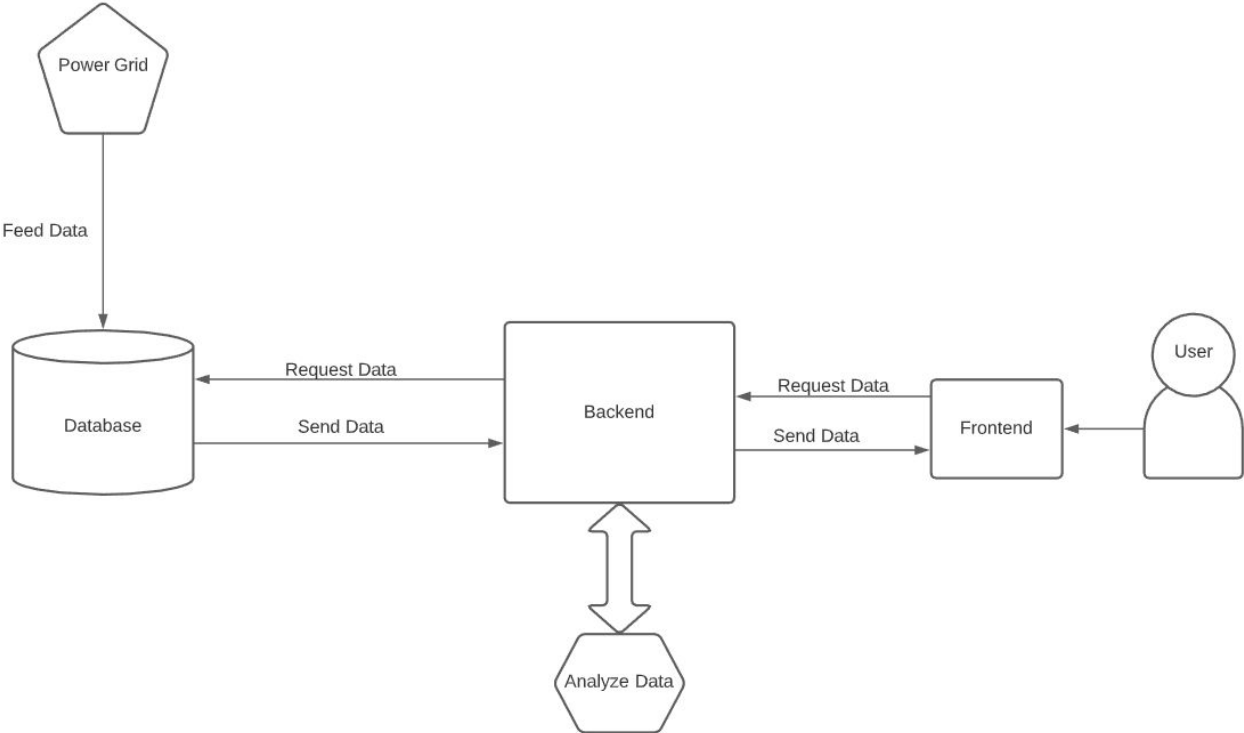
Abhilash Tripathy

# Project Vision

- The Grid AI project seeks to use Deep Machine Learning to develop insights and analytics on a power grid in real time.



# Conceptual Design Diagram



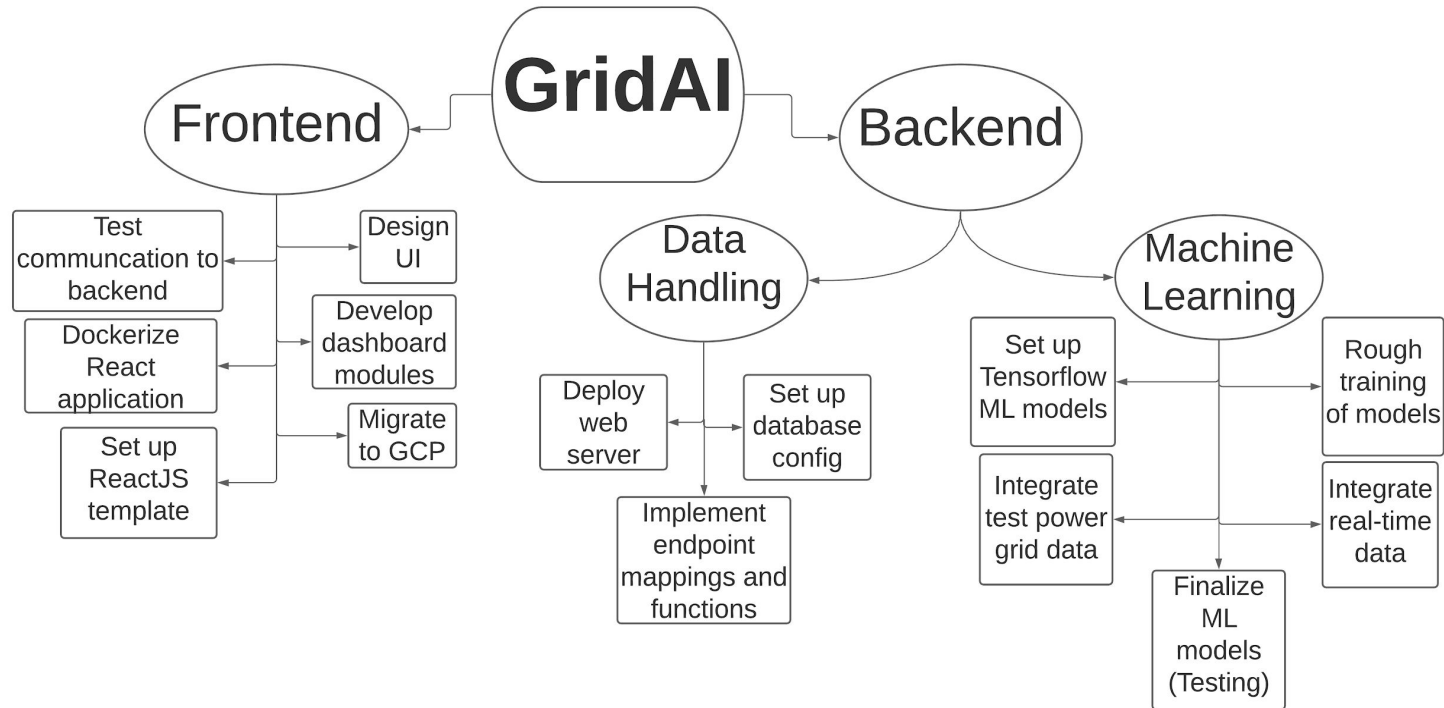
# Requirements: Functional

- Analyze grid data in real-time
  - Neo4j
  - Flask
- Run on remote system
  - Google Cloud Platform
- Accurately predict and classify anomalies
  - Tensorflow/Keras
- Visualize data and analysis in comprehensible format
  - React (Material UI)

# Requirements: Non-Functional

- Scalability
  - Docker
  - Modular Design
- Maintainable
  - Loosely coupled
- Performance
  - Lightweight backend framework

# Planned Tasks Overview



# Risks and Mitigation

Risks:

Machine Learning:

- Real-world power grids have several variables to consider
  - Size of Power Grid
  - Population of Consumers affecting power draw
  - Demand in different seasons
- The accuracy of the machine learning algorithm to detect anomalies will only be trained on OPAL-RT

Cloud Integration:

- \$300 GCP access to be utilized later.
- Deployment issues on server will be difficult to resolve with time constraint.

# Risks and Mitigation

Mitigation:

Machine Learning:

- Larger pool of data; try different ML algorithms to test with the DNN.

GCP setup

- Use trial benefits to understand the platform and requirements.

Task related issues - Priority Rating

- Resolve higher priority issues first.

Test Components according to IEEE standards:

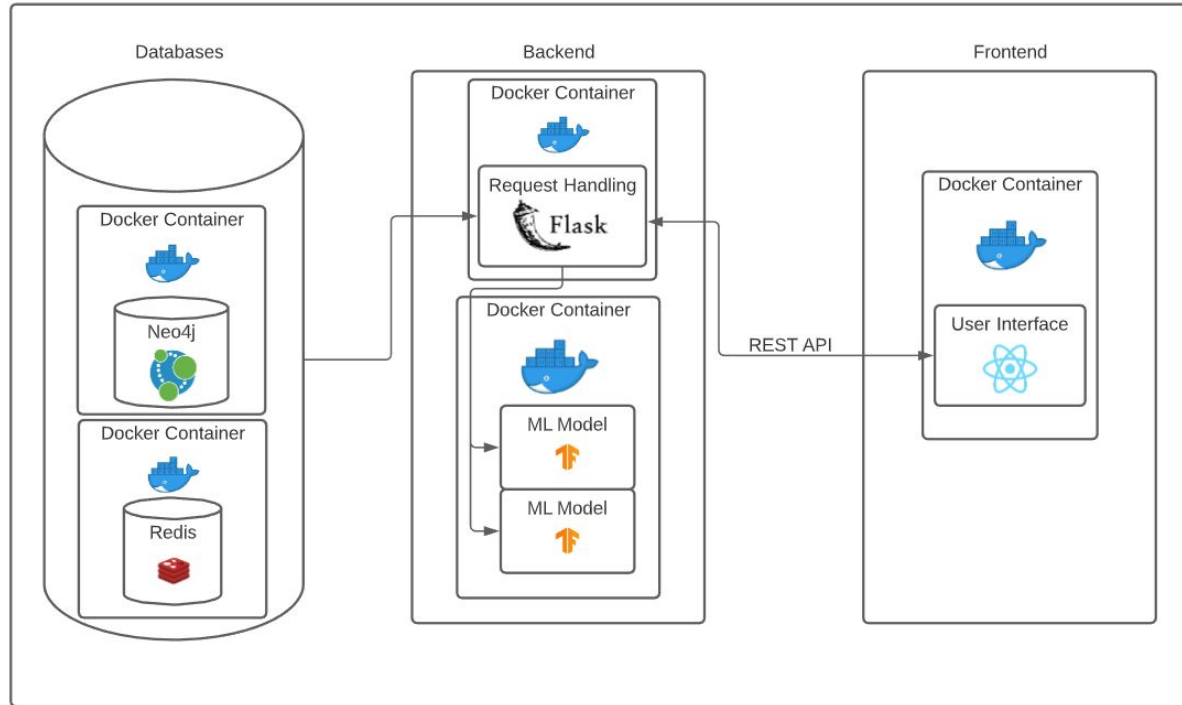
- Evaluate component functionality and reliability with others (check modularity).



# System Design: Architecture

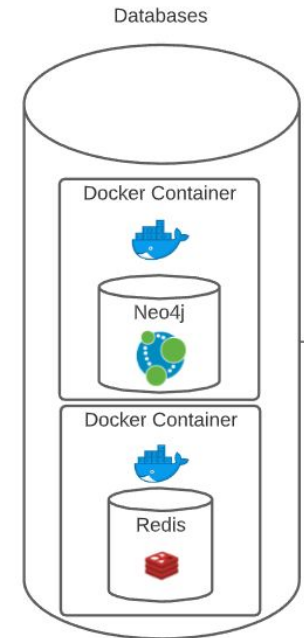


Google Cloud  
Google Cloud  
Platform



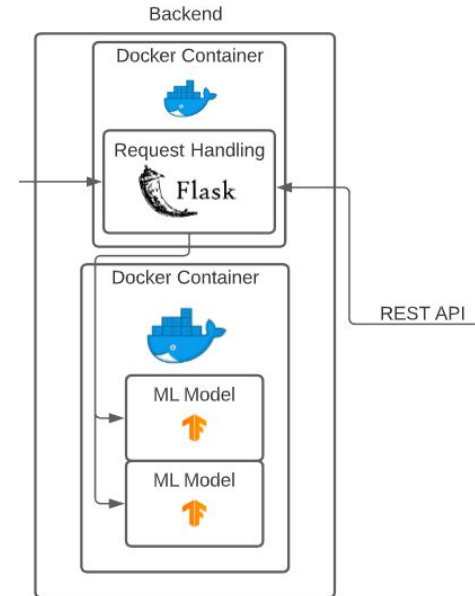
# System Design: Databases

- Neo4j
  - Main Database
  - NoSQL node architecture
  - Store grid data
- Redis
  - Cache Memory Database
  - NoSQL architecture
  - Potentially store ML analysis for short time



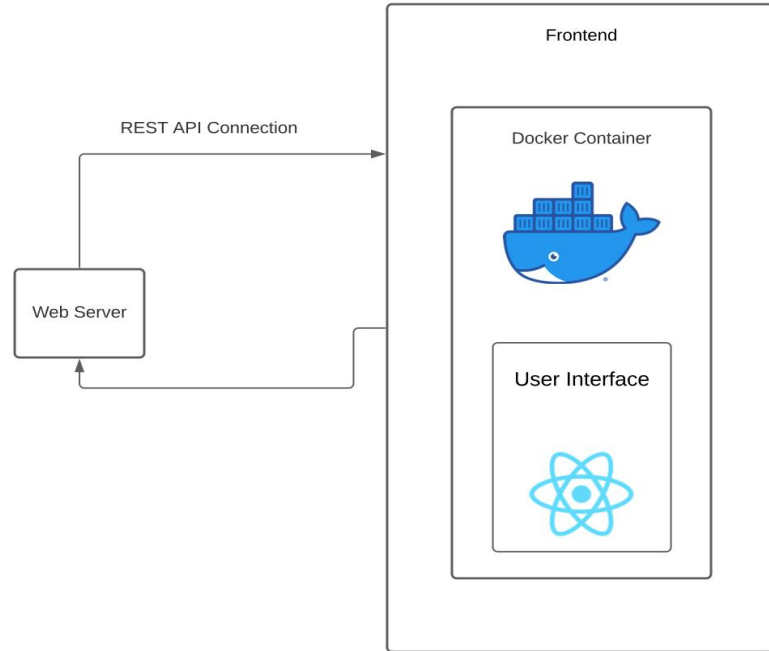
# System Design: Backend

- Request Handling
  - Flask
    - Lightweight
  - REST API
  - JSON data structure
- Machine Learning Models
  - Tensorflow & Keras
  - Predictor and Classifier
  - Create more as necessary



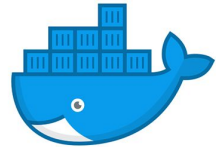
# System Design: Frontend

- User Interface
  - Web Application
  - React
    - Material UI



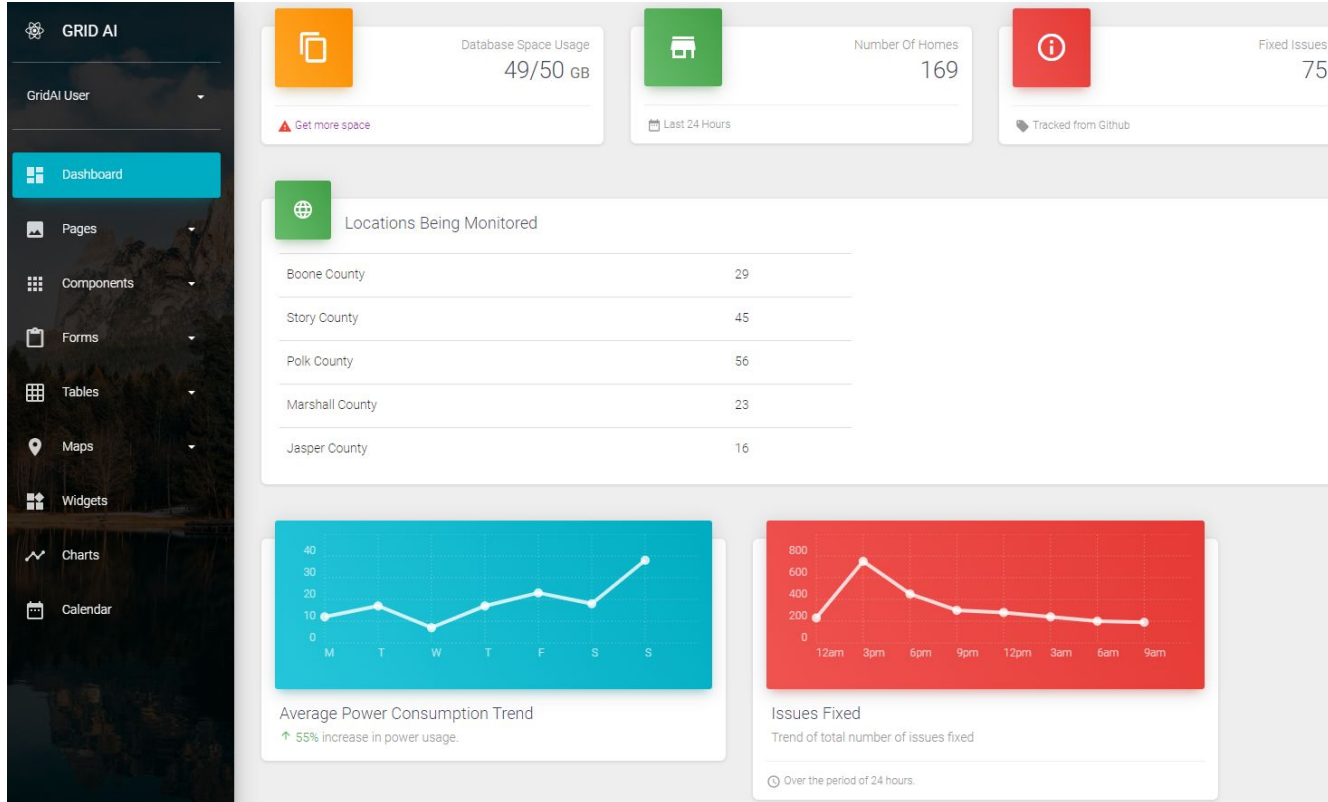
# System Design: Deployment

- Docker
  - Containerize individual components
  - Allows for efficient deployment
  - Do not have to worry about host system configuration
- ISU PowerCyber Testbed
  - Developmental environment
- Google Cloud Platform
  - Cheap and accessible
  - Reduce resources needed from our end and user end



Google Cloud

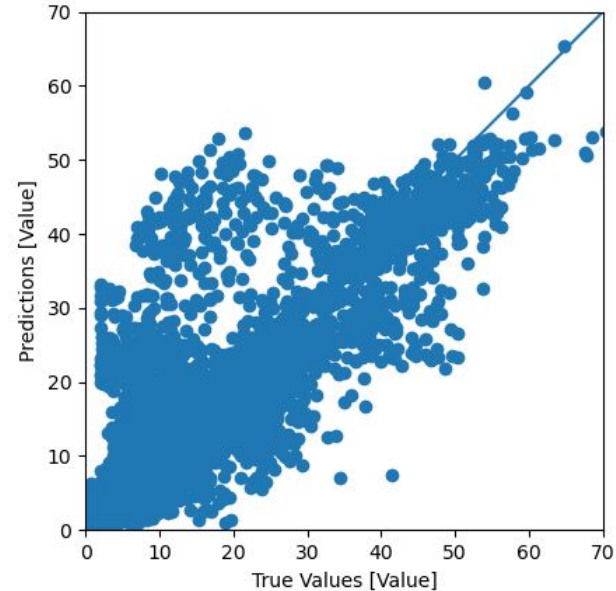
# Frontend Prototype



# Machine Learning Prototype

- Initial naive implementation for estimating node outputs
- Currently takes static node information and timestamp as input into DNN
- Does not include output from previous nodes in chain

Plot of Prediction vs Actual Value For Feeder A



# Project Plan: Milestones

Milestone Duration and Progression Metrics:

- Difficulty of tasks based on team-member experience
- Client evaluation
- Meeting functional and non-functional criterias
- Weightage based on estimated sub tasks



# Project Plan: Milestones

GridAI Timeline

sdmay21-23

	Weeks 1-3	Weeks 4-6	Weeks 7-9	Weeks 10-12	Weeks 13-15
<i>Setup React template</i>	Frontend				
<i>Deploy web server</i>	Backend				
<i>Setup database configuration</i>	Backend	Backend			
<i>Develop ML algorithm</i>	Backend	Backend	Backend		
<i>Develop UI for dashboard</i>	Frontend	Frontend	Frontend		
<i>Train ML algorithm</i>			Backend	Backend	
<i>Mapping of URLs</i>		Backend	Backend		
<i>Integrate real-time data</i>		Both	Both	Both	
<i>Dockerize for PowerCyber testbed</i>	Frontend	Frontend			
<i>Test frontend communication</i>			Frontend	Frontend	Frontend
<i>Test ML Accuracy</i>				Backend	Backend
<i>Test system integration</i>				Both	Both



# Test Plan: ML Models

- 3-stage Process
  - Predicting output using the training data set
  - Predicting output using the OpenDSS simulator on the same parameters as the training set
  - Predicting output of a generic power grid using Opal-RT
- At each stage the expected and actual values will be compared to be acceptable by the client.

# Test Plan: Backend Functions

- Utilize Postman
  - Test individual endpoints with dummy data
  - Test database integration
  - Test ML model integration

# Test Plan: Frontend Functions & Interfaces

- Manual Testing
  - Making sure data is getting processed in JavaScript functions correctly
  - Validate data showing in graphs
  - Verify components that need perform functions perform them and perform them correctly
- Unit Testing
  - Jest
- Acceptance Testing
  - Verify no components overlap, run off screen, or aren't showing up
  - Have faculty advisor/client also validate our interface design

# Conclusion

- Still in the early stages of project implementation
  - Machine Learning Model (Justin Merkel, Karthik Prakash, Abir Mojumder)
    - Docker Container Configured
    - Development in progress
  - Backend (Abir Mojumder, Karthik Prakash, Justin Merkel)
    - Database Setup in progress
    - REST API endpoint mappings in progress
  - Frontend (Patrick Wenzel, Abhilash Tripathy)
    - UI template implemented

# Next Semester Plans

## Backend

- Set up Google Cloud Platform (GCP) account
- Set up database instances in GCP
- Develop and train machine and deep learning models
- Set up data pipeline

## Frontend

- Develop frontend interface
- Connect frontend to GCP database
- Developing ability to set up queries on the frontend to get data from the backend
  - Verify data on the frontend side
  - Visualize the data

Questions?