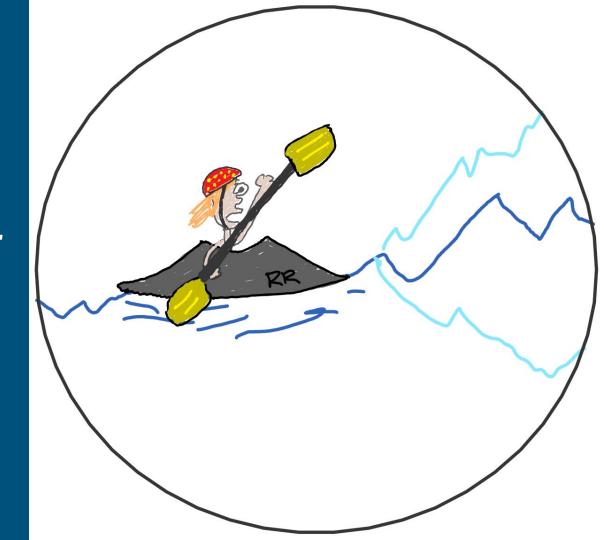
River Runner

Whitewater Kayaking Predictions

Ryan Bald, Kenten Danas, Luke Waninger



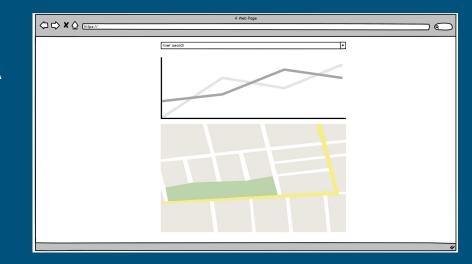
Background

- Whitewater kayaking is a popular sport in western Washington
- But, rivers are highly dynamic. Flow rates can change drastically in short time periods.
- There are currently no reliable tools to predict a river's flow rate, meaning paddlers are often left to figure out whether they can run a given river less than a day in advance





- River Runner aims to provide predictions for whitewater kayaking run flow rates up to a week in advance
- Predictions are made based on ARIMA models of past flow rates along with exogenous weather predictors
- Paddlers will be able to access predictions for a chosen run(s) via a publically available graphical user interface



Data Sources: Professor Paddle



- Used for:
 - River run data including river run names and max/min runnable flow rates
- Obtained by:
 - Single use scraping functions
- Limitations:
 - This site does not have max/min runnable flow rates for all runs



Data Sources: **USGS**



- Used for:
 - Historical and ongoing streamflow data
- Obtained by:
 - Repeated calls to the USGS Instantaneous Values REST web service
- Limitations:
 - Restricts the amount of data returned
 - Not every station measured streamflow

Data Sources: NOAA



- Used for:
 - Historical exogenous predictor data
 - snowpack, precipitation, temperature, humidity
- Obtained by:
 - Manual requests submitted through NOAA's climate data online portal
- Limitations
 - Restricted how much data could be accessed with a single manual request
 - Data was provided via an email link and could not be automated for ongoing retrieval

Data Sources: <u>DarkSky</u>



- Used for:
 - Ongoing exogenous predictor data
 - precipitation, temperature, humidity
- Obtained by:
 - Continuous API calls to DarkSky (https://api.darksky.net/forecast)
- Limitations
 - Data is retrieved for the previous day, every day, and works as expected. We found no limitations in regards to our project.

Use Cases



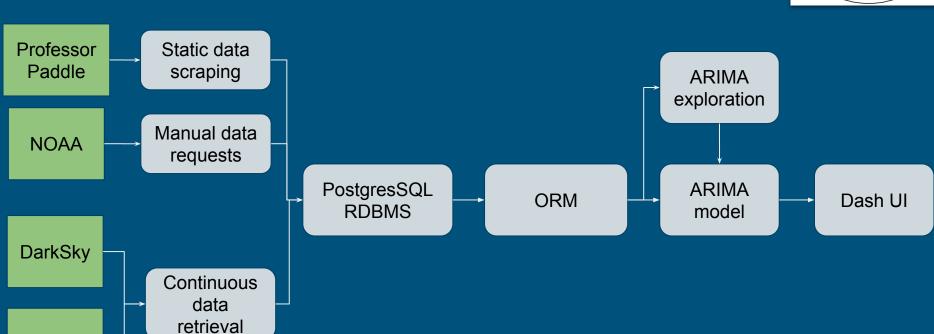
- Paddlers use RiverRunner to help plan kayaking trips by getting a prediction of a selected river's flow rate, which provides an indication of whether the river will be runnable
- User accesses the tool through a webpage, and selects a river run by name from the drop-down filter or by data point on the map
- The system responds with a plot for that run which includes:
 - 3 weeks of historic flow rate
 - 7 days of predicted flow rate
 - Highlighted band between the maximum and minimum runnable flow rates (where available)

Demo



Design

USGS



Design - Ongoing Data Collection



- Daily script that retrieves all weather/river-related data from the previous day and computes predictions
- Script that can retrieve all weather/river-related data over a specified date

range if need be

```
def daily_run():
    context = Context(settings.DATABASE)
    session = context.Session()

    get_weather_observations(session)
    get_usgs_observations()
    compute_predictions(session)

    session.close()
```

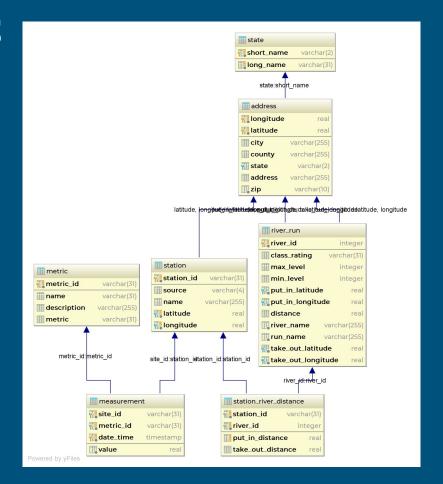
```
predictions = arima.arima_model(run.run_id)

to_add = [
    Prediction(
        run_id=run.run_id,
        timestamp=pd.to_datetime(d),
        fr_lb=round(float(p), 1),
        fr=round(float(p), 1),
        fr_ub=round(float(p), 1)
    )
    for p, d in zip(predictions.values, predictions.index.values)
]

repo.clear_predictions(run.run_id)
repo.put_predictions(to_add)
log(f'predictions for {run.run_id}-{run.run_name} added to db')
```

Design - RDBMS

 Persistence is managed through an RDBMS -PostgresSQL 10.3 - Ubuntu Server 16.04 LTE

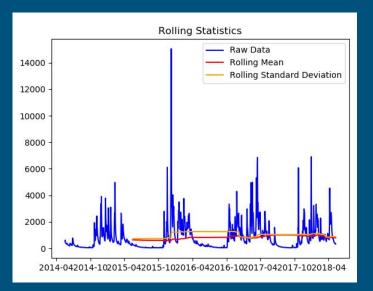


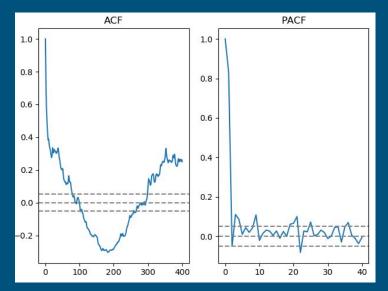






- Static module used once for data exploration to determine best model parameters
- Analysis included visual and statistical (Dickey-Fuller) tests for stationarity,
 autocorrelation analysis for AR order and partial autocorrelation analysis for MA order





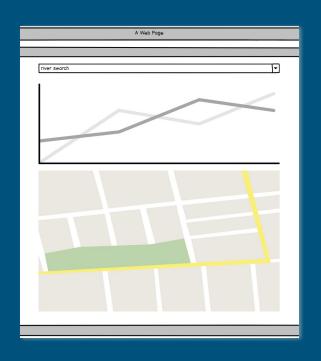
Design - ARIMA Model

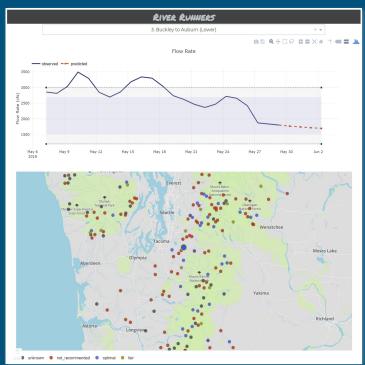


- Future river flow rates are predicted using an autoregressive integrated moving average (ARIMA) model generated from historic USGS river flow rate time series data
- Temperature and precipitation are included in the models as exogenous predictor variables
- Models are generated using the past four years of historical data up to the current day, and predictions are made for the future seven days.
- Modelling is completed using built in Python functions

Design - Dash Front End







CI Testing - TeamCity



TC build

success

Project Structure

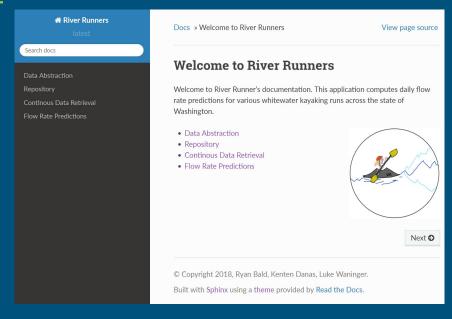


High Level

- RiverRunner/
 - o doc/
 - o examples/
 - riverrunner/
 - tests/
 - o .gitignore
 - LICENSE
 - o README.md

https://github.com/kentdanas/RiverRunne

<u>r</u>



Lessons Learned & Future Work



Lessons Learned

- Getting one ARIMA model to converge for 300 different datasets is difficult
- Dash is a quick and effective tool for web visualization
- Database must be prepped for daily modeling

Future Work

- Further refine the models used to improve accuracy and extend the forecasted flow rates beyond 7 days
- Integrate weather forecast data for exogenous predictors
- Expand scope of project to include more river runs.

Questions?

