AI-VVO sdmay22-36 Weekly Update #7

11/8/2021 - 11/15/2021

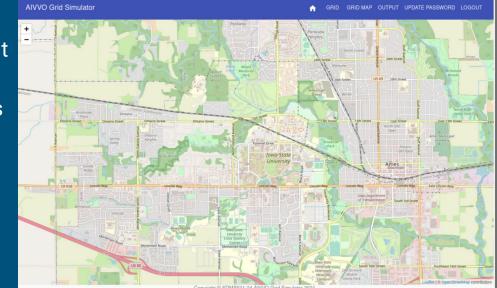
- Completed more react-leaflet tutorials
- Began implementing map features into project in our VMs
- Researched markers and custom icons
- Installed react-leaflet and leaflet into the project
- Researched using a json data file for getting node data to map

- Implemented a Map.js file to output the map on the application
- Will eventually implement the Map.js file into the GridVis.js file for less confusion

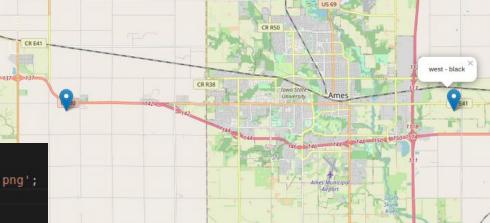
const position = [42.03, -93.65];

```
<MapContainer className="map"
    center={position}
    zoom={10}
    style={{ height: 750, width: "100%" }}
>
<TileLayer
    attribution='&amp;copy <a href="http://url="https://{s}.tile.openstreetmap.org/>
/>
</MapContainer>
```

- Added Grid map button to the home screen toolbar (will eventually change it to be in the Grid screen toolbar)
- The button will display the map with its center being at a specified location



- Fixed issue with markers not displaying by importing leaflet marker icons
- Set up a default icon to use



import icon from 'leaflet/dist/images/marker-icon.png'; import iconShadow from 'leaflet/dist/images/marker-shadow.png';

```
let DefaultIcon = L.icon({
    iconUrl: icon,
    shadowUrl: iconShadow
});
```

- Added a test "data.json" file with data from a tutorial
- JSON file will be filled with node data including:
 - Coordinates
 - Status (online, offline, etc.)
 - Last Date of Repair
 - Etc.
- Updated README with tutorial information and helpful links

```
end \geq react_app \geq src \geq data \geq {} data.json \geq [ ] features \geq {} 25
```

```
"type": "FeatureCollection",
"crs": {
"type": "name",
"properties": { "name": "urn:ogc:def:crs:OGC:1.3:CRS84" ]
}.
```

```
"features": [
```

```
"properties": {
 "PARK ID": 960,
 "FACILITYID": 28014.
 "NAME": "Bearbrook Skateboard Park",
 "NAME FR": "Planchodrome Bearbrook",
 "ADDRESS": "8720 Russell Road"
 "ADDRESS FR": "8720, chemin Russell",
 "FACILITY T": "flat".
 "FACILITY 1": "plat",
 "ACCESSIBLE": "no/non",
 "NOTES": "Outdoor"
 "MODIFIED D": "2018/01/18",
 "CREATED DA": null,
 "FACILITY": "Neighbourhood : smaller size facility to service population of 10,000 or less",
 "FACILITY F": "De voisinage : petite installation assurant des services à 10 000 résidents ou moins.",
 "DESCRIPTIO": "Flat asphalt surface, 5 components".
 "DESCRIPT 1": "Surface d'asphalte plane, 5 modules",
 "PICTURE LI": null.
 "PICTURE DE": null,
 "PICTURE 1": null
"geometry": {
 "coordinates": [-75.3372987731628, 45.383321536272049]
```

Front-end (Next Week)

- Put Map and associated libraries in a Docker container
- Populate JSON file with test data based on real nodes
- Place markers according to real node locations on map
- Research how to add route highlighting for connected networks

• Created two new tables in the database for FeederAP and FeederAQ data

List of relations			
Schema	Name	Type	Owner
+		++	
public	auth_group	table	postgres_user
public	auth_group_permissions	table	postgres_user
public	auth_permission	table	postgres_user
public	auth_user	table	postgres_user
public	auth_user_groups	table	postgres_user
public	auth_user_user_permissions	table	postgres_user
public	authtoken_token	table	postgres_user
public	django_admin_log	table	postgres_user
public	django_content_type	table	postgres_user
public	django_migrations	table	postgres_user
public	django_session	table	postgres_user
public	prediction_feeder_ap_data	table	postgres_user
public	prediction_feeder_aq_data	table	postgres_user
(13 rows)			

A

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• Filled in the tables with data from csv files given

	id date_time bus_1 bus_2 bus_3 bus_4 bus_5 bus_6 bus_7 bus_8 b us_9 bus_10 bus_11 bus_12 bus_13 bus_14 bus_15 bus_16 bus_17
p data	0 1/1/2017 1:00 0.0000000 0.0000000 15.2900000 6.8920000 4.9160000 5.0400000 4.1630000 14.0960000 17.
	0810000 7.1360000 3.1250000 3.0830000 1.5370000 0.4580000 0.7930000 3.0830000 2.1310000 1 1/1/2017 2:00 0.0000000 0.0000000 14.9010000 6.6720000 5.3350000 4.7600000 3.0700000 14.9370000 12.
	7860000 7.0780000 2.7580000 2.0320000 2.3780000 0.3360000 0.8360000 2.0320000 1.9140000 2 1/1/2017 3:00 0.00000000 0.0000000 15.7720000 7.0130000 4.5630000 5.0400000 3.5070000 14.7890000 10. 2090000 5.9910000 3.0960000 1.5970000 1.6150000 0.3260000 0.830000 1.5970000 2.4460000
	3 1/1/2017 4:00 0.0000000 1.5770000 15.770000 6.4520000 4.7820000 4.8000000 3.1430000 14.7610000 10. 0400000 7.0300000 3.3170000 1.2280000 1.5360000 0.4220000 0.7780000 1.2280000 6.0410000
	4 1/1/2017 5:00 0.0000000 0.0000000 15.2920000 6.3560000 4.4820000 5.0000000 3.1470000 15.1560000 10. 1470000 6.0430000 2.8320000 1.1620000 1.7730000 0.4040000 0.8740000 1.1620000 4.8800000
	5 1/1/2017 6:00 0.0000000 0.0000000 15.8140000 6.8610000 4.9630000 4.3600000 3.3360000 11.1450000 9. 6780000 6.0750000 4.4330000 1.7980000 1.6460000 0.3500000 0.8170000 1.7980000 3.6560000 6 1/1/2017 7:00 0.0000000 0.0000000 16.0440000 8.4220000 4.7000000 5.2400000 3.3200000 9.6230000 9.
	3270000 5.7060000 4.3310000 1.0500000 1.5880000 0.4150000 0.8540000 1.0500000 2.0050000 7 1/1/2017 8:00 0.0000000 0.0000000 15.3370000 8.2010000 4.6640000 4.1200000 3.5720000 9.3930000 9.
	5300000 5.8090000 7.2480000 1.6700000 1.7150000 0.8280000 1.0100000 <u>1.6700000 2.3810000</u> id date_time bus_1 bus_2 bus_3 bus_4 bus_5 bus_6 bus_7 bus_8 b
	us_9 bus_10 bus_11 bus_12 bus_13 bus_14 bus_15 bus_16 bus_17 +
g data	0 1/1/17 1:00 AM 0.00000000 0.00000000 3.8320352 3.3379479 1.9429275 2.2962918 2.0162329 6.4223273 6. 7508432 2.8203277 1.0271378 1.1189796 0.4482917 0.1951072 0.1610255 1.4931650 0.6215417
quata	1 1/1/17 2:00 AM 0.0000000 0.0000000 4.8977219 3.0398530 2.4306978 0.9665592 0.4374513 5.9034801 3. 7292500 1.4372492 0.5600358 0.9841425 1.1517180 0.1530861 0.1697570 0.9841425 0.6946892
	2 1/1/17 3:00 AM 0.0000000 0.0000000 5.1840057 2.0454583 1.8034130 1.4700000 0.7121267 7.1626396 2. 5586166 0.8536713 0.7759307 0.6803192 0.4047572 0.1485300 0.2914501 0.2275602 1.0419916
	3 1/1/17 4:00 ÅM 0.00000000 0.00000000 7.1791013 3.1248462 1.3947500 2.0447914 1.3389124 5.3575276 4. 8625939 2.3106493 1.5112698 0.4036241 0.4480000 0.1922689 0.2823763 0.3581667 2.1925902
	4 1/1/17 5:00 AM 0.00000000 0.00000000 5.0262373 1.8538333 1.4731622 0.7124614 1.0343689 2.1596130 4. 6231098 2.7532721 0.5750621 0.4592518 0.8078027 0.0575669 0.3454269 0.2912247 1.2230433
	5 1/1/17 6:00 AM 0.0000000 0.0000000 4.6124167 1.7195287 1.9615031 1.8573522 0.6774037 4.0450948 1. 3790403 1.7718750 1.2929583 0.8191930 0.2345423 0.1270330 0.2685349 0.6525869 1.5574495
	6 1/1/17 7:00 AM 0.0000000 0.0000000 5.2734078 4.0789608 1.3708333 2.0709805 1.0912312 1.9540335 3. 6862663 0.8130610 1.2632083 0.5085382 0.2262777 0.1210417 0.1216884 0.4783941 0.9135050
	7 1/1/17 8:00 AM 0.0000000 0.0000000 4.4732917 3.9719256 1.1689086 1.0325693 0.5089824 2.3541077 4. 3419963 2.8134271 3.5103666 0.8088179 0.7305869 0.1681326 0.3319709 0.5489025 0.4834827

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Django model for the tables

Aq model looks the same

class feeder ap data(models.Model): date time = models.TextField() bus 1 = models.DecimalField(max digits=11, decimal places=7) bus 2 = models.DecimalField(max digits=11, decimal places=7) bus 3 = models.DecimalField(max digits=11, decimal places=7) bus 4 = models.DecimalField(max digits=11, decimal places=7) bus 5 = models.DecimalField(max digits=11, decimal places=7) bus 6 = models.DecimalField(max digits=11, decimal places=7) bus 7 = models.DecimalField(max digits=11, decimal places=7) bus 8 = models.DecimalField(max digits=11, decimal places=7) bus 9 = models.DecimalField(max digits=11, decimal places=7) bus 10 = models.DecimalField(max digits=11, decimal places=7) bus 11 = models.DecimalField(max digits=11, decimal places=7) bus 12 = models.DecimalField(max digits=11, decimal places=7) bus 13 = models.DecimalField(max digits=11, decimal places=7) bus 14 = models.DecimalField(max digits=11, decimal places=7) bus 15 = models.DecimalField(max digits=11, decimal places=7) bus 16 = models.DecimalField(max digits=11, decimal places=7) bus 17 = models.DecimalField(max digits=11, decimal places=7)

 Created a CSV parser script that takes the csv file and directly inputs it into the postgreSQL database

```
print('input path of csv file to add:')
csvFileToAdd = input()
print('input database table name:')
databaseTableName = input()
columnNames = ('date time', )
for i in range(1, 18):
    columnNames += ('bus ' + str(i), )
dataframe = pd.read csv(csvFileToAdd, header=0, names=columnNames)
print(dataframe)
print('does this data look correct?(y/n)')
confirmAddition = input()
if confirmAddition == 'n':
    quit()
#connect to db
engine = sqlalchemy.create engine('postgresgl://postgres user:postgres password@l
dataframe.to sql(databaseTableName, engine, index label='id', if exists='append')
print('table added successfully')
```

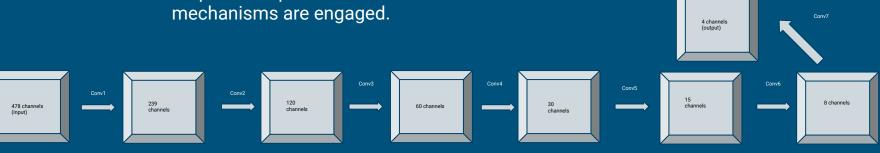
Back-end (Next Week)

- Decide if the feeder data works well in the database table. And if it does add the Feeders B and C into datatables
- Work on the connection between ML algorithm and backend api that will send the data to the front end

- Implemented AIVVONet class on the PowerCyberTestBed and pushed class file to Gitlab
- Extended nn.Module to construct AIVVONet
- Structure of neural network is as follows:
 - 478 inputs per time step
 - 7 convolution layers with each layer reducing the size of the problem by a factor of 2
 - 4 Control Mechanisms, so the output will represent what control mechanisms are engaged.

```
import numpy as np
import torch
import torch.nn as nn
import torch.optim as optimizer
class AIVVONet(nn.Module):
    def __init__(self):
        self.conv2 = nn.Conv2d(in channels = 239, out channels = 120
         self.conv3 = nn.Conv2d(in_channels = 120, out_channels = 60)
         self.conv4 = nn.Conv2d(in channels = 60, out channels = 30)
         self.conv5 = nn.Conv2d(in channels = 30, out channels = 15)
        self.conv6 = nn.Conv2d(in channels = 15, out channels = 8)
         self.conv7 = nn.Conv2d(in channels = 8, out channels = 4)
    def forward(self,x):
         x = nn.functional.relu(self.conv1(x))
        x = nn.functional.relu(self.conv2(x))
        x = nn.functional.relu(self.conv3(x))
        x = nn.functional.relu(self.conv4(x))
        x = nn.functional.relu(self.conv5(x))
        x = nn.functional.relu(self.conv6(x))
        x = nn.functional.relu(self.conv7(x))
        return x
```

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10 11

23 24

25 26

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- Installed PyTorch on the PowerCyberTestbed VM
 - Installed pip3 with "apt install python3-pip"
 - Installed PyTorch dependencies with "pip3 install torch torchvision torchaudio"

```
root@ubuntu-vm:/home/ubuntu# pip3 install torch torchvision torchaudio
Collecting torch
 Downloading torch-1.10.0-cp38-cp38-manylinux1 x86 64.whl (881.9 MB)
                                      | 881.9 MB 3.5 kB/s
Collecting torchvision
 Downloading torchvision-0.11.1-cp38-cp38-manylinux1_x86_64.whl (23.3 MB)
                                       23.3 MB 64.8 MB/s
Collecting torchaudio
 Downloading torchaudio-0.10.0-cp38-cp38-manylinux1 x86 64.whl (2.9 MB)
                                      1 2.9 MB 49.2 MB/s
Collecting typing-extensions
 Downloading typing extensions-4.0.0-pv3-none-anv.whl (22 kB)
Requirement already satisfied: pillow!=8.3.0.>=5.3.0 in /usr/lib/python3/dist-pa
ckages (from torchvision) (7.0.0)
Collecting numpy
 Downloading numpy-1.21.4-cp38-cp38-manylinux 2 12 x86 64.manylinux2010 x86 64.
whl (15.7 MB)
                                      | 15.7 MB 19.2 MB/s
Installing collected packages: typing-extensions, torch, numpy, torchvision, tor
chaudio
Successfully installed numpy-1.21.4 torch-1.10.0 torchaudio-0.10.0 torchvision-0
.11.1 typing-extensions-4.0.0
root@ubuntu-vm:/home/ubuntu#
```

- Installed all required components of Django on VM-3
- Worked with Django to develop an understanding of how it interacts with PostgreSQL to store and retrieve data
- Create a simple test "view" to read and display data from out tables.
- Added our two models to the Django Admin page to view and manipulate for easy testing.

Feeder AP
11/14/2021
1.000000
Feeder AQ
11/14/2021
27.000000

```
 Feeder AP 
{% for ap data in ap.all %}
     {{ap data.date time}}
    <br>>
    {{ap data.bus 1}}
    <hr>>
{% endfor %}
<br>>
 Feeder AQ 
<br>>
{% for aq data in aq.all %}
     {{aq data.date time}}
    <br>>
    \langle p \rangle  {{aq data.bus 1}}\langle p \rangle
    <hr>>
{% endfor %}
```

Home > Prediction > Feeder ag datas AUTH TOKEN The feeder ag data "feeder ag data object (1)" was added successfully Tokens + Add Select feeder_aq_data to change AUTHENTICATION AND AUTHORIZATION Action: _____ ✓ Go 0 of 1 selected Groups + Add Users + Add FEEDER AO DATA feeder_aq_data object (1) PREDICTION 1 feeder ag data Feeder_ap_datas + Add Feeder_aq_datas + Add # Retreive Data

```
def Pull_Data_Test(request):
```

```
#Get data as an Object for Feeder AP
ap_list = feeder_data.feeder_ap_data.objects
```

```
#Get data as an Object for Feeder AQ
aq_list = feeder_data.feeder_aq_data.objects
```

#Utilize the render shortcut to display data
return render(request, "tests.html", {'ap' : ap list,'aq':aq list})

Machine Learning (Next Week)

- GridController agent class will be implemented
- Main learning loop utilizing GridController and AIVVONet will be constructed
- Add PyTorch and apt install python3-pip to the Docker install files
- Modify the test view to be a more complete data retrieval function
- Use the data retrieval function to provide data to the main Machine Learning Algorithm.