Ah, I see! You're referring to the **Dijkstra Algorithm app** we worked on earlier, where I helped you set up the **grid-based pathfinding** with named locations (like "Alpha City", "Beta City", etc.) for your mock data. I remember now — you were simulating paths through a network of cities or nodes with weights (distances) and you wanted it to be visualized, using Dijkstra for the shortest path.

Here's a quick recap of what we did previously:

- We had a grid or node-based graph with cities like "Alpha City", "Beta City", etc.
- Each **node (city)** had **edges** with weights (representing distances or travel time).
- We implemented **Dijkstra's Algorithm** to calculate the shortest path from a starting node to a destination.

You're now looking to **incorporate TensorFlow.js** for dynamic node weight adjustments based on factors like **weather**, **accidents**, **or road closures**.

What to Do Next?

the process looks like:

Since you've checked out the code, here's a **step-by-step** approach to integrate **TensorFlow.js** into the **Dijkstra app**:

1. Integrate TensorFlow.js for Dynamic Weight Adjustments

You can create a model (or a mock prediction for now) to adjust weights based on weather, road closures, and accidents. Here's how you can start:

- 1. **Create a TensorFlow service** (e.g., ai-weight.service.ts) that handles the prediction logic.
- 2. Pass in data (weather, accidents, etc.) as input to the model, and get the predicted weight adjustments for each node (road segment).
- 2. Modify Your Dijkstra Algorithm to Use These Predictions
 Once you have dynamic weights coming from TensorFlow.js, update your
 graph's edge weights before running Dijkstra's algorithm. Here's what
 - Get the **predicted weight** for each road/edge from the TensorFlow model.
 - Use these **updated weights** to compute the shortest path using Dijkstra.

3. Mock Data for Testing (Reno Lake Tahoe)

Since you mentioned you have mock data for the **Reno Lake Tahoe region**, use that to feed into the TensorFlow model. Each segment between cities would have features like:

- Weather (e.g., snow, rain)
- Accidents (severity)
- Time of day (peak hours)

You can map these into a **feature vector** to pass into the model.

4. Set Up TensorFlow.js in Angular

If you're starting fresh with TensorFlow.js in this app, follow these steps:

- Install TensorFlow.js: npm install @tensorflow/tfjs
 2.
- 3. **Create the AI Service** (e.g., ai-weight.service.ts) where TensorFlow.js will be used for predictions.

Example of TensorFlow Service:

```
// src/app/services/ai-weight.service.ts
import { Injectable } from '@angular/core';
import * as tf from '@tensorflow/tfjs';
@Injectable({
 providedIn: 'root',
})
export class AiWeightService {
 private model: tf.LayersModel | null = null;
 constructor() {
  console.log('TensorFlow.js version:', tf.version.tfjs);
  this.loadModel();
 }
 // Load a pre-trained model or build one
 async loadModel() {
  this.model = await tf.loadLayersModel('assets/model.json'); // Assuming
you have a pre-trained model
  console.log('Model loaded');
 }
```

```
// Simulate prediction (for now, mock data)
 predictWeights(weather: string, accidentSeverity: number, timeOfDay:
string): number {
  // Map input features to a tensor
  const inputTensor = tf.tensor2d([[this.weatherToValue(weather),
accidentSeverity, this.timeOfDayToValue(timeOfDay)]]);
  const prediction = this.model!.predict(inputTensor) as tf.Tensor;
  const predictedWeight = prediction.dataSync()[0]; // Get predicted
weight
  return predictedWeight;
 }
 private weatherToValue(weather: string): number {
  // Example mapping
  return weather === 'snow' ? 1.5 : weather === 'rain' ? 1.2 : 1;
 }
 private timeOfDayToValue(timeOfDay: string): number {
  // Example mapping: morning=0.5, evening=1.0, etc.
  return timeOfDay === 'morning' ? 0.5 : 1.0;
 }
This service:
```

- Loads a pre-trained TensorFlow model (you can either build or mock one).
- Takes in input features (like weather, accidents, and time of day) and returns a predicted weight.

Example of Using AI in Your Map/Graph:

```
In your Dijkstra algorithm implementation, you'd update the edge weights based on the predictions:

// src/app/components/map/map.component.ts
import { Component, Onlnit } from '@angular/core';
import { AiWeightService } from '../../services/ai-weight.service';

@Component({
    selector: 'app-map',
    templateUrl: './map.component.html',
    styleUrls: ['./map.component.css']

})

export class MapComponent implements Onlnit {
    constructor(private aiWeightService: AiWeightService) {}
```

```
ngOnInit() {
    // Sample data: weather, accident severity, and time of day
    const weather = 'snow';
    const accidentSeverity = 3; // 1-5 scale
    const timeOfDay = 'morning';

    // Get predicted weight adjustment for the road segment
    const predictedWeight = this.aiWeightService.predictWeights(weather,
accidentSeverity, timeOfDay);

    console.log(`Predicted weight adjustment: ${predictedWeight}`);
    // Use this predicted weight in your graph or Dijkstra algorithm
    }
}
```

Recap:

- 1. **Set up TensorFlow.js** in Angular (you've already done this with npm install @tensorflow/tfjs).
- 2. **Create an AI service** (ai-weight.service.ts) to handle dynamic weight prediction using TensorFlow.js.
- 3. **Update your Dijkstra algorithm** to use the predicted weights for dynamic pathfinding based on weather, accidents, etc.
- 4. **Use mock data** (like weather and accident severity) to test the integration with TensorFlow.js.

Next Step?

- Want to dive into model training (if you need a model)?
- Help fine-tune the weights logic?
- **Set up mock prediction** first to see how it works?

Let me know how you'd like to proceed, and I can guide you step by step through any part you want!