

```
!pip install osmnx
```

```
Collecting osmnx
```

```
  Downloading osmnx-2.0.2-py3-none-any.whl.metadata (4.9 kB)
```

```
Requirement already satisfied: geopandas>=1.0 in
```

```
/usr/local/lib/python3.11/dist-packages (from osmnx) (1.0.1)
```

```
Requirement already satisfied: networkx>=2.5 in
```

```
/usr/local/lib/python3.11/dist-packages (from osmnx) (3.4.2)
```

```
Requirement already satisfied: numpy>=1.22 in
```

```
/usr/local/lib/python3.11/dist-packages (from osmnx) (2.0.2)
```

```
Requirement already satisfied: pandas>=1.4 in
```

```
/usr/local/lib/python3.11/dist-packages (from osmnx) (2.2.2)
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```
Requirement already satisfied: requests>=2.27 in
```

```
/usr/local/lib/python3.11/dist-packages (from osmnx) (2.32.3)
```

```
Requirement already satisfied: shapely>=2.0 in
```

```
/usr/local/lib/python3.11/dist-packages (from osmnx) (2.1.0)
```

```
Requirement already satisfied: pyogrio>=0.7.2 in
```

```
/usr/local/lib/python3.11/dist-packages (from geopandas>=1.0->osmnx)  
(0.10.0)
```

```
Requirement already satisfied: packaging in
```

```
/usr/local/lib/python3.11/dist-packages (from geopandas>=1.0->osmnx)  
(24.2)
```

```
Requirement already satisfied: pyproj>=3.3.0 in
```

```
/usr/local/lib/python3.11/dist-packages (from geopandas>=1.0->osmnx)  
(3.7.1)
```

```
Requirement already satisfied: python-dateutil>=2.8.2 in
```

```
/usr/local/lib/python3.11/dist-packages (from pandas>=1.4->osmnx)  
(2.8.2)
```

```
Requirement already satisfied: pytz>=2020.1 in
```

```
/usr/local/lib/python3.11/dist-packages (from pandas>=1.4->osmnx)  
(2025.2)
```

```
Requirement already satisfied: tzdata>=2022.7 in
```

```
/usr/local/lib/python3.11/dist-packages (from pandas>=1.4->osmnx)  
(2025.2)
```

```
Requirement already satisfied: charset-normalizer<4,>=2 in
```

```
/usr/local/lib/python3.11/dist-packages (from requests>=2.27->osmnx)  
(3.4.1)
```

```
Requirement already satisfied: idna<4,>=2.5 in
```

```
/usr/local/lib/python3.11/dist-packages (from requests>=2.27->osmnx)  
(3.10)
```

```
Requirement already satisfied: urllib3<3,>=1.21.1 in
```

```
/usr/local/lib/python3.11/dist-packages (from requests>=2.27->osmnx)  
(2.3.0)
```

```
Requirement already satisfied: certifi>=2017.4.17 in
```

```
/usr/local/lib/python3.11/dist-packages (from requests>=2.27->osmnx)  
(2025.1.31)
```

```
Requirement already satisfied: six>=1.5 in
```

```
/usr/local/lib/python3.11/dist-packages (from python-dateutil>=2.8.2->  
pandas>=1.4->osmnx) (1.17.0)
```

```
Downloading osmnx-2.0.2-py3-none-any.whl (99 kB)
```

99.9/99.9 kB 3.3 MB/s eta

0:00:00

nx

Successfully installed osmnx-2.0.2

```
# Importing all necessary libraries
```

```
import osmnx as ox
```

```
import networkx as nx
```

```
import matplotlib.pyplot as plt
```

```
import matplotlib.colors as mcolors
```

```
import matplotlib.cm as cm
```

```
import random
```

```
import numpy as np
```

```
import pandas as pd
```

```
from collections import defaultdict
```

```
class RoadNetwork:
```

```
    """
```

```
    Represents a road network loaded from OpenStreetMap using OSMnx.
```

```
    """
```

```
    def __init__(self, address, distance=1000, network_type='drive'):
```

```
        """
```

```
        Initializes the RoadNetwork with data from OSMnx.
```

```
        Args:
```

```
            address (str): Address to center the network around.
```

```
            distance (int, optional): Radius (in meters) around the  
address to load. Defaults to 1000.
```

```
            network_type (str, optional): Type of network to load  
(e.g., 'drive', 'walk'). Defaults to 'drive'.
```

```
        Raises:
```

```
            ValueError: If network cannot be loaded for the given  
address.
```

```
        """
```

```
        try:
```

```
            self.graph = ox.graph_from_address(address, dist=distance,  
network_type=network_type)
```

```
            # Check if graph is empty
```

```
            if len(self.graph.nodes) == 0:
```

```
                raise ValueError("Loaded network contains no nodes")
```

```
            if len(self.graph.edges) == 0:
```

```
                raise ValueError("Loaded network contains no edges")
```

```
            self._add_travel_time_attribute()
```

```
        except Exception as e:
```

```
            raise ValueError(f"Failed to load network for address  
'{address}': {str(e)}")
```

```

def _add_travel_time_attribute(self):
    """
    Adds a 'travel_time' attribute to each edge based on speed
    limit and length.
    Handles different speed formats and units.

    Formula: travel_time (minutes) = length (meters) / (speed
    (km/h) * 1000/60)
    Where:
        - length is in meters
        - speed is in kilometers per hour
        - 1000/60 converts km/h to m/min
    """
    for u, v, k, data in self.graph.edges(data=True, keys=True):
        # Get maxspeed or default to 30 km/h
        speed = data.get("maxspeed", "30")

        # Handle list of speeds
        if isinstance(speed, list):
            speed = speed[0]

        # Clean and convert speed value
        speed_value = 30.0 # Default value

        if isinstance(speed, str):
            # Extract numeric part and unit
            speed_parts = speed.strip().split()
            speed_num = speed_parts[0]
            try:
                speed_value = float(speed_num)

                # Convert mph to km/h if necessary
                if len(speed_parts) > 1 and "mph" in
speed_parts[1].lower():
                    speed_value *= 1.60934 # Convert mph to km/h
            except (ValueError, TypeError):
                print(f"Warning: Could not parse speed value
'{speed}', using default of 30 km/h")
            elif isinstance(speed, (int, float)):
                speed_value = float(speed)

        # Calculate travel time (in minutes) with explicit unit
        # conversion
        length_m = data["length"] # length in meters
        speed_kmh = speed_value # speed in km/h
        meters_per_minute = (speed_kmh * 1000) / 60 # Convert
        km/h to m/min
        data["travel_time"] = length_m / meters_per_minute #
        Result in minutes

```

```

        # Calculate road capacity based on road properties
        lanes = data.get("lanes", 1)
        if isinstance(lanes, list):
            lanes = lanes[0]
        try:
            lanes = float(lanes)
        except (ValueError, TypeError):
            lanes = 1.0

        road_type = data.get("highway", "residential")
        # Base capacity depends on road type
        if road_type in ["motorway", "trunk", "primary"]:
            base_capacity = 25 # Higher capacity for major roads
        elif road_type in ["secondary", "tertiary"]:
            base_capacity = 15 # Medium capacity
        else:
            base_capacity = 8 # Lower capacity for minor roads

        data["capacity"] = base_capacity * lanes

    def plot(self):
        """
        Plots the road network using OSMnx.

        Returns:
            tuple: Figure and Axes objects of the plot.
        """
        fig, ax = ox.plot_graph(self.graph, figsize=(10,10))
        return fig, ax

class Car:
    """
    Represents a car in the traffic simulation.
    """
    def __init__(self, start, destination, graph=None):
        """
        Initializes the Car with a start and destination node.

        Args:
            start (int): Node ID of the starting location.
            destination (int): Node ID of the destination location.
            graph (networkx.MultiDiGraph, optional): The road network
graph.

        Attributes:
            current_location (int): Current node ID where the car is
located.
            destination (int): Target node ID where the car is
heading.
            path (list): Ordered list of node IDs representing the

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planned route,
                                excluding the current location (first node is
the next node to visit).
                                graph (networkx.MultiDiGraph, optional): Reference to the
road network graph.
    """
    if graph is not None:
        # Validate nodes exist in graph
        if start not in graph:
            raise ValueError(f"Start node {start} does not exist in
the graph")
        if destination not in graph:
            raise ValueError(f"Destination node {destination} does
not exist in the graph")

    self.current_location = start
    self.destination = destination
    self.path = []
    self.graph = graph # Reference to the graph

    # Initialize path if graph is provided
    if graph is not None:
        self.calculate_path(graph)

    def calculate_path(self, graph):
        """
        Calculates the shortest path from current location to
destination.

        Args:
            graph (networkx.MultiDiGraph): The road network graph.

        Returns:
            bool: True if path was found, False otherwise.
        """
        try:
            import networkx as nx
            full_path = nx.shortest_path(graph, self.current_location,
self.destination, weight="travel_time")
            self.path = full_path[1:] # Exclude current location
            return True
        except (nx.NetworkXNoPath, nx.NodeNotFound):
            return False

    def recalculate_path_with_congestion(self, graph, congestion_map,
congestion_threshold=0.7): # Corrected indentation
        """
        Recalculates path avoiding congested roads.
        """
        import networkx as nx

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```

# Create a copy of the graph with congestion weights
G_temp = graph.copy()

# Apply congestion penalties to edge weights
for u, v, k, data in G_temp.edges(data=True, keys=True):
    congestion = congestion_map.get((u, v, k), 0)
    if congestion > congestion_threshold:
        # Penalize congested roads by increasing travel time
        congestion_penalty = 1 + (congestion * 5) # Up to 6x
longer
        G_temp[u][v][k]["travel_time_adjusted"] =
data["travel_time"] * congestion_penalty
    else:
        G_temp[u][v][k]["travel_time_adjusted"] =
data["travel_time"]

    try:
        # Find path with adjusted weights
        return nx.shortest_path(G_temp, self.current_location,
self.destination,
                                weight="travel_time_adjusted")
    except (nx.NetworkXNoPath, nx.NodeNotFound):
        # Fall back to original path if no alternative found
        return nx.shortest_path(graph, self.current_location,
self.destination,
                                weight="travel_time")

class TrafficSimulator:
    """
    Simulates traffic flow on a road network.
    """
    def __init__(self, road_network, num_cars=100, start_hour = 8):
        """
        Initializes the TrafficSimulator with a road network and
        number of cars.

        Args:
            road_network (RoadNetwork): The road network to simulate
on.
            num_cars (int, optional): Number of cars in the
simulation. Defaults to 100.
        """
        self.road_network = road_network
        self.num_cars = num_cars
        self.cars = self._create_cars()
        self.current_hour = start_hour
        self.minutes = 0
        self.minutes_per_step = 1 # Define minutes_per_step (I decided
to set at 1 as it gave interesting results)

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        self.congestion_map = {}

    def _create_cars(self):
        """
        Creates cars with random starting and destination locations.
        Initializes their paths using the road network.
        """
        nodes = list(self.road_network.graph.nodes())
        cars = []

        for _ in range(self.num_cars):
            start = random.choice(nodes)
            destination = random.choice([n for n in nodes if n !=
start])
            car = Car(start, destination, self.road_network.graph)
            # Ensure car has a valid path
            if not car.path and car.current_location != car.destination:
                try:
                    car.path = nx.shortest_path(self.road_network.graph,
                                                car.current_location,
                                                car.destination,
                                                weight="travel_time")
[1:] # Exclude current location
                except nx.NetworkXNoPath:
                    # Try a different destination
                    continue
            cars.append(car)

        return cars

    def move_cars(self):
        """
        Moves cars in the simulation based on their paths and traffic
        conditions.
        """
        # Update congestion map periodically
        if not hasattr(self, 'step_counter'):
            self.step_counter = 0
        self.step_counter += 1

        if self.step_counter % 5 == 0:
            self.update_congestion_map()

        # Track if any car moved for debugging
        any_car_moved = False

        for car in self.cars:
            # Check if car has reached its destination
            if car.current_location == car.destination:
                # Generate a new destination

```

```

        nodes = list(self.road_network.graph.nodes())
        car.destination = random.choice([node for node in nodes if
node != car.current_location])
        try:
            car.path = nx.shortest_path(self.road_network.graph,
                                       car.current_location,
                                       car.destination,
                                       weight="travel_time")[1:] #
Exclude current location
        except nx.NetworkXNoPath:
            continue # Skip this car if no path found
            # print(f"Car reached destination, new path length:
{len(car.path)}")
            continue

        # Skip cars with empty paths
        if not car.path:
            # Try to recalculate path
            try:
                car.path = nx.shortest_path(self.road_network.graph,
                                           car.current_location,
                                           car.destination,
                                           weight="travel_time")[1:]
                #print(f"Recalculated path, new length:
{len(car.path)}")
            except nx.NetworkXNoPath:
                # If no path, assign new destination
                nodes = list(self.road_network.graph.nodes())
                car.destination = random.choice([node for node in
nodes if node != car.current_location])
                try:
                    car.path =
nx.shortest_path(self.road_network.graph,
                                           car.current_location,
                                           car.destination,
                                           weight="travel_time")
[1:]
                # print(f"New destination, path length:
{len(car.path)}")
            except nx.NetworkXNoPath:
                continue

            if not car.path:
                continue

        next_node = car.path[0]

        # Check if the edge exists (in case the graph changed)
        if next_node not in
self.road_network.graph[car.current_location]:

```



```

        # Path is invalid, recalculate
        try:
            car.path = nx.shortest_path(self.road_network.graph,
                                       car.current_location,
                                       car.destination,
                                       weight="travel_time")[1:]

        except nx.NetworkXNoPath:
            continue
        if not car.path:
            continue
        next_node = car.path[0]

        # Find the edge data
        edge_data = None
        for k in self.road_network.graph[car.current_location]
[next_node]:
            edge_data = self.road_network.graph[car.current_location]
[next_node][k]
            break

        # Count cars on edge
        cars_on_edge = sum(1 for c in self.cars if c.current_location
== car.current_location and c.path and c.path[0] == next_node)

        # Calculate movement probability based on congestion
        capacity = edge_data.get('capacity', 8)

        # Apply time factors
        time_factor = 1.0
        if 7 <= self.current_hour <= 9 or 16 <= self.current_hour <=
18: # RUsh hour
            time_factor = 0.6

        effective_capacity = capacity * time_factor
        congestion_factor = min(1.0, cars_on_edge /
effective_capacity)
        speed_reduction = 1.0 - (congestion_factor * 0.8)

        # ONLY ONE movement decision per car per step
        if random.random() < speed_reduction:
            car.current_location = next_node
            car.path.pop(0)
            any_car_moved = True

        # Dynamic rerouting (separate from movement)
        reroute_probability = 0.05
        if congestion_factor > 0.7:
            reroute_probability = 0.3

        if random.random() < reroute_probability and hasattr(car,

```

```

'recalculate_path_with_congestion'):
    try:
        new_path = car.recalculate_path_with_congestion(
            self.road_network.graph,
            self.congestion_map
        )
        if new_path and len(new_path) > 1:
            car.path = new_path[1:]
    except Exception as e:
        print(f"Error in rerouting: {e}")

    if not any_car_moved:
        print("WARNING: No cars moved this step!")

def simulate(self, num_steps=10):
    """
    Runs the traffic simulation for a given number of steps.

    Args:
        num_steps (int, optional): Number of simulation steps.
    Defaults to 10.
    """
    for step in range(num_steps):
        # Update simulation time
        self.minutes += self.minutes_per_step
        if self.minutes >= 60:
            self.current_hour += (self.current_hour + 1) % 24
            self.minutes %= 60

        # Move cars - this MUST be properly indented to run every
        step
        self.move_cars()
        #print(f"Step {step + 1} - Time: {self.current_hour:02d}:
        {self.minutes:02d}")

        # Check if anything is happening
        car_moved = False
        for car in self.cars:
            if car.path:
                car_moved = True
                break

        if not car_moved:
            print("WARNING: No cars are moving in the simulation!")

        # For demonstration, print the number of cars at each node
        """
        node_counts = {node: 0 for node in
        self.road_network.graph.nodes()}
        for car in self.cars:

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        node_counts[car.current_location] += 1
    print(f"Step {step + 1}:")
    for node, count in node_counts.items():
        print(f"Node {node}: {count} cars")
    print()
    """

def plot_traffic(self):
    """
    Improved traffic visualization using congestion (cars /
    capacity).
    Adds colorbar and better styling.
    """
    # Update congestion map before plotting
    self.update_congestion_map()
    G = self.road_network.graph

    # Get node positions for layout
    node_positions = {node: (data['x'], data['y']) for node, data
in G.nodes(data=True)}

    # Boost low congestion values
    congestion_boost = {k: v**0.5 for k, v in
self.congestion_map.items()} # gamma correction
    max_congestion = max(congestion_boost.values()) if
congestion_boost else 1
    norm = mcolors.Normalize(vmin=0,
vmax=max(congestion_boost.values()))
    cmap = cm.get_cmap("OrRd")

    fig, ax = plt.subplots(figsize=(10, 10))
    fig.patch.set_facecolor('lightgray') # Set figure
background
    ax.set_facecolor('lightgray') # Set plot (axes)
background

    for (u, v, k), congestion in self.congestion_map.items():
        color = cmap(norm(congestion))
        width = 2 + 4 * congestion # line thickness increases
with congestion

        x_vals = [node_positions[u][0], node_positions[v][0]]
        y_vals = [node_positions[u][1], node_positions[v][1]]
        ax.plot(x_vals, y_vals, color=color, linewidth=width,
alpha=0.9)

    # Draw nodes
    nx.draw_networkx_nodes(G, node_positions, ax=ax, node_size=5,
node_color="gray", alpha = 0.7)

```

```

    # Colorbar legend
    sm = plt.cm.ScalarMappable(cmap=cmap, norm=norm)
    sm.set_array([])
    cbar = plt.colorbar(sm, ax=ax, shrink=0.7, pad=0.01)
    cbar.set_label("Congestion Level (cars / capacity)",
fontsize=12)
    # Add ticks with readable values
    cbar_ticks = np.linspace(0, max(congestion_boost.values()),
num=6)
    cbar.set_ticks(cbar_ticks)
    cbar.set_ticklabels([f"{tick:.2f}" for tick in cbar_ticks])
    cbar.ax.yaxis.set_tick_params(color='black',
labelcolor='black')
    plt.setp(plt.getp(cbar.ax.axes, 'yticklabels'), color='black')

    ax.set_title("Traffic Congestion Map", fontsize=15,
color="black")
    ax.tick_params(left=False, bottom=False, labelleft=False,
labelbottom=False, colors = 'white')
    plt.axis("off")
    plt.show()

    return fig, ax

def _edge_car_counts(self):
    """
    Calculates the number of cars on each edge of the road
network.

    Returns:
        dict: A dictionary mapping edge keys (u, v, k) to car
counts.
    """
    counts = {}
    for u, v, k in self.road_network.graph.edges(keys=True):
        counts[(u, v, k)] = 0 # Using u, v, k as key

    for car in self.cars:
        if car.path:
            next_node = car.path[0]
            # Find the correct edge key
            for k in self.road_network.graph[car.current_location]
[next_node].keys():
                counts[(car.current_location, next_node, k)] += 1
            break # Assuming we only want to count one edge
between the nodes
    return counts

def update_congestion_map(self):
    """

```

```

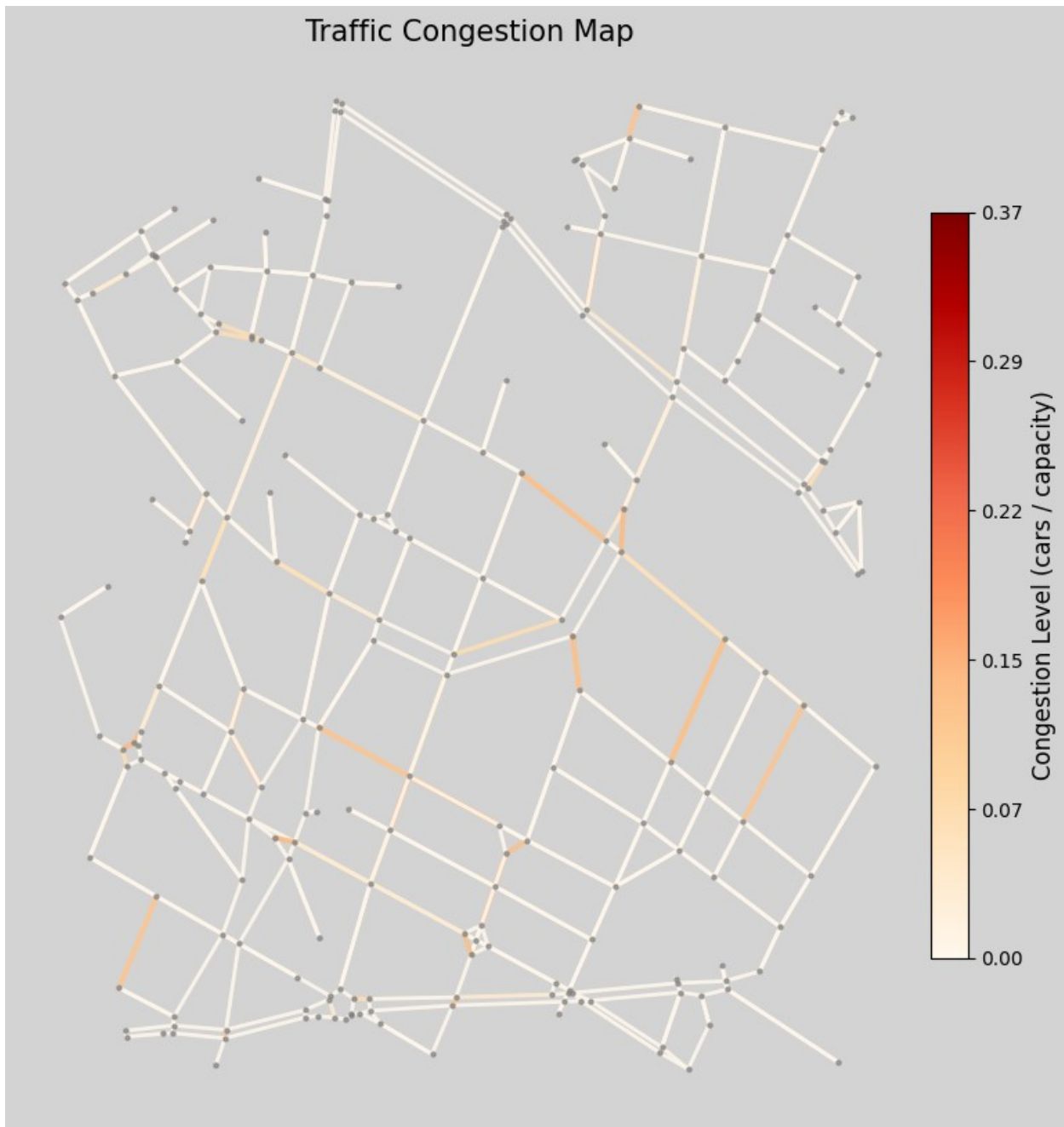
Creates a map of congested edges in the network.
"""
self.congestion_map = {}

# Calculate congestion for each edge
for u, v, k, data in self.road_network.graph.edges(data=True,
keys=True):
    cars_on_edge = sum(1 for car in self.cars
                        if car.current_location == u
                        and car.path
                        and car.path[0] == v)
    capacity = data.get('capacity', 8) # Get capacity from edge
    congestion_level = cars_on_edge / capacity
    self.congestion_map[(u, v, k)] = congestion_level

berlin_network = RoadNetwork('Adalbertstraße 58, Berlin, Germany')
simulator = TrafficSimulator(berlin_network)
simulator.simulate()
simulator.plot_traffic()

<ipython-input-6-5b8c658043ff>:220: MatplotlibDeprecationWarning: The
get_cmap function was deprecated in Matplotlib 3.7 and will be removed
in 3.11. Use ``matplotlib.colormaps[name]`` or
``matplotlib.colormaps.get_cmap()`` or ``pyplot.get_cmap()`` instead.
cmap = cm.get_cmap("OrRd")

```



```
(<Figure size 1000x1000 with 2 Axes>,
 <Axes: title={'center': 'Traffic Congestion Map'}>)

## Simulating Berlin network for 10 times
# a. Empirical Analysis
num_simulations = 10 # Number of simulations to run
congestion_data = [] # Store congestion levels for each simulation

for _ in range(num_simulations):
    simulator = TrafficSimulator(berlin_network)
```

```

    simulator.simulate(num_steps=100) # Run simulation for 100 steps
    in each loop
    congestion_data.append(simulator.congestion_map)

# Analyze congestion_data to identify consistently congested edges
# For example, calculate average congestion for each edge:
edge_congestion_avg = {}
for edge in berlin_network.graph.edges(keys=True):
    edge_congestion_avg[edge] = np.mean([d.get(edge, 0) for d in
congestion_data])

# Print the most congested edges
# This will show which edges are most congested across simulations
import pprint
pprint.pprint(edge_congestion_avg)

simulator.plot_traffic()

{(21487224, 29215073, 0): np.float64(0.006666666666666666),
(21487224, 196725581, 0): np.float64(0.003333333333333333),
(21487230, 26960762, 0): np.float64(0.003333333333333333),
(21487230, 28794539, 0): np.float64(0.0),
(21487230, 196724115, 0): np.float64(0.006666666666666666),
(21487230, 1491865195, 0): np.float64(0.0),
(21487231, 29215047, 0): np.float64(0.013333333333333332),
(21487231, 5984003368, 0): np.float64(0.0),
(21487232, 26960762, 0): np.float64(0.006666666666666666),
(21487232, 29218293, 0): np.float64(0.02),
(21487232, 29218315, 0): np.float64(0.01),
(21487232, 269476782, 0): np.float64(0.004444444444444444),
(26517409, 287447548, 0): np.float64(0.002222222222222222),
(26960758, 29271261, 0): np.float64(0.006666666666666666),
(26960761, 28794542, 0): np.float64(0.019999999999999997),
(26960761, 29218316, 0): np.float64(0.0125),
(26960761, 2625899861, 0): np.float64(0.02),
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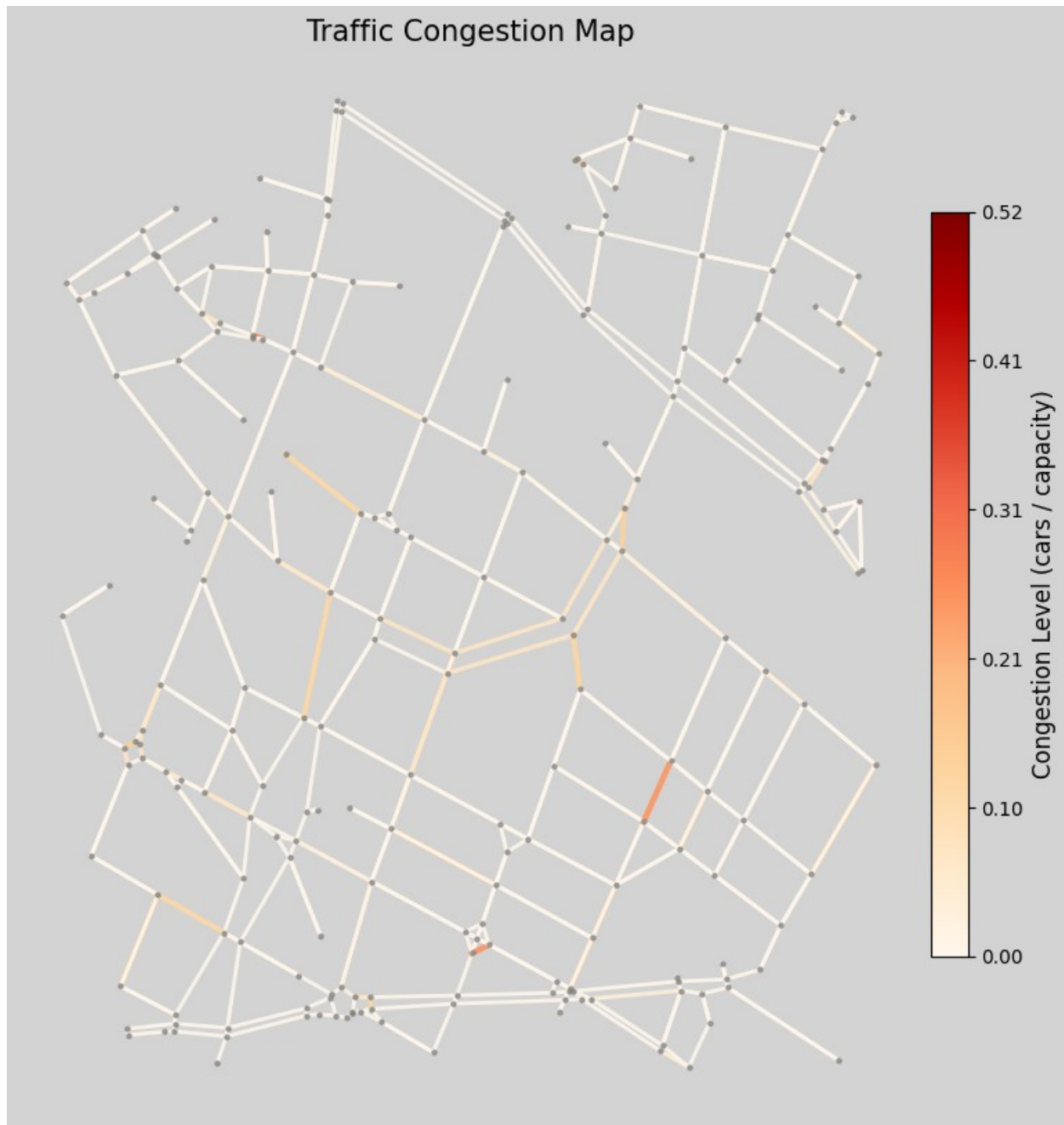
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in 3.11. Use ``matplotlib.colormaps[name]`` or  
``matplotlib.colormaps.get_cmap()`` or ``pyplot.get_cmap()`` instead.  
cmap = cm.get_cmap("OrRd")
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(<Figure size 1000x1000 with 2 Axes>,
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Theoretical Analysis

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betweenness centrality =
nx.betweenness centrality(berlin_network.graph, weight="travel_time")
edge_betweenness centrality =
nx.edge_betweenness centrality(berlin_network.graph,
weight="travel_time")
closeness centrality = nx.closeness centrality(berlin_network.graph,
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```

distance="travel_time")
degree centrality = nx.degree_centrality(berlin_network.graph)
print(f"Between centrality:", betweenness_centrality)
print(f"Edge between centrality:", edge_betweenness_centrality)
print(f"Closeness centrality:", closeness_centrality)
print(f"Degree centrality:", degree_centrality)

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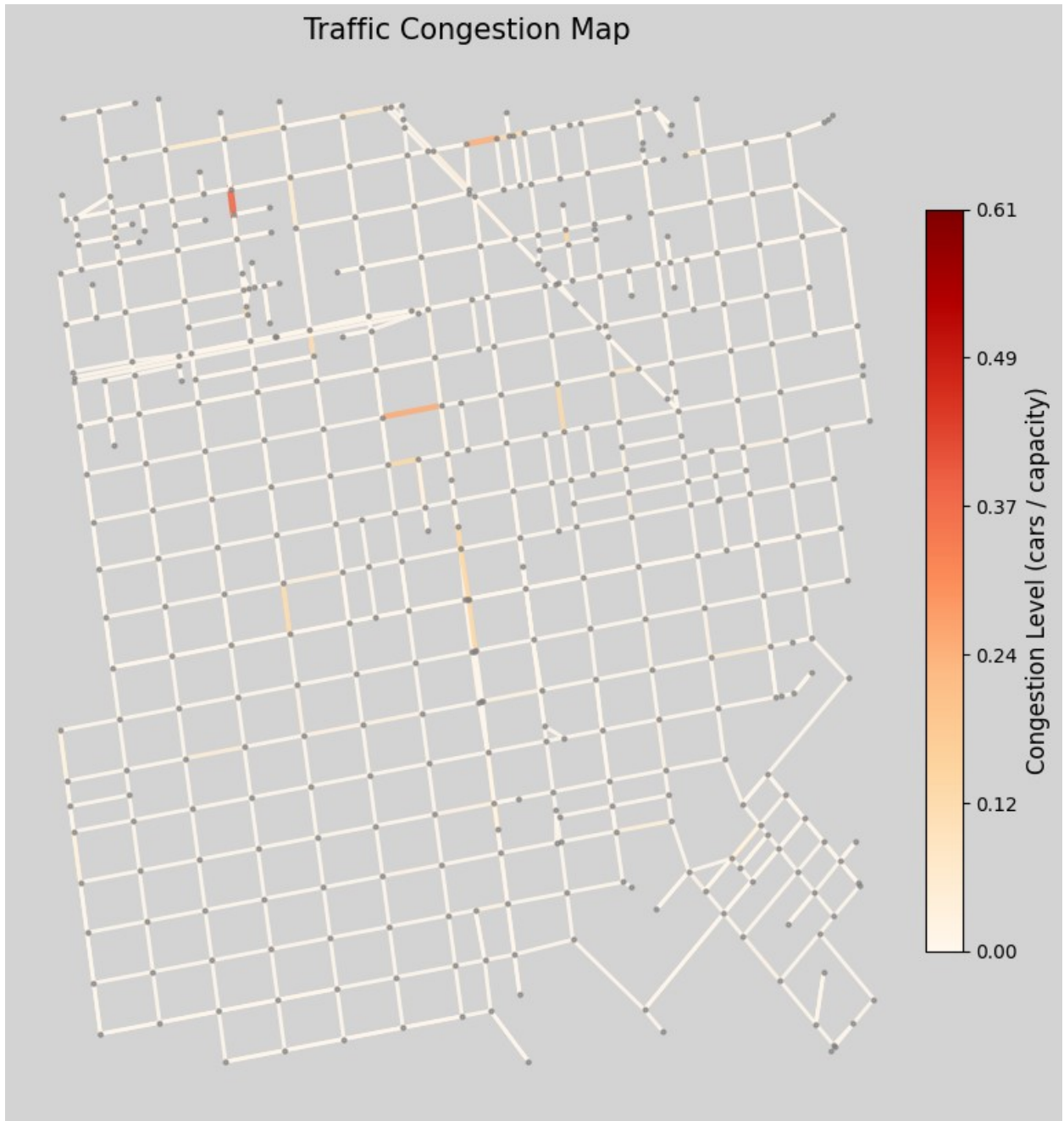
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sf_network = RoadNetwork('California 851, San Francisco, United
States')
```

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simulator = TrafficSimulator(sf_network)
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```
simulator.simulate(num_steps = 100)
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```
simulator.plot_traffic()
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<ipython-input-6-5b8c658043ff>:220: MatplotlibDeprecationWarning: The
get_cmap function was deprecated in Matplotlib 3.7 and will be removed
in 3.11. Use ``matplotlib.colormaps[name]`` or
``matplotlib.colormaps.get_cmap()`` or ``pyplot.get_cmap()`` instead.
cmap = cm.get_cmap("OrRd")
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(<Figure size 1000x1000 with 2 Axes>,  
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