

# Sumo×PyPSA: Interactive Web Demo of Real-Time Urban Power-Traffic Co-Simulation with Vehicle-to-Grid

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## Abstract

We demonstrate Sumo×PyPSA, an interactive browser-based web platform for exploring real-time bidirectional coupling between urban power grids and traffic networks. Users may interact with a live visualization of Manhattan's infrastructure—8 substations, 15 transmission lines, 3,481 traffic signals, and dynamic vehicle traffic—to trigger cascading blackouts, deploy emergency Vehicle-to-Grid (V2G) responses, and observe bidirectional dynamics. The demonstration enables hands-on experimentation with four scenarios: (1) summer heatwave causing cascading failures across multiple substations, (2) automated V2G emergency response preventing blackouts by routing vehicles to provide 400 kW grid support, (3) manual substation failure with operator-controlled recovery, and (4) custom scenario creation. The browser-based interface runs on commodity hardware, accessible from any device without specialized software installation, and is implemented using standard web technologies (Flask, WebSocket, and Mapbox GL JS). Validated against real Manhattan data (OpenStreetMap, NYC DOT, DOE CBECS), the platform demonstrates that complex infrastructure co-simulation can be made accessible to policymakers, educators, and researchers through intuitive web technologies.

**Demo video:** <https://xgraph-team.github.io/SumoXPypsa/website/><sup>1</sup>

## Keywords

Interactive demonstration, Web applications, Power grids, Traffic networks, Vehicle-to-Grid, Co-simulation

## 1 Introduction

Urban infrastructure systems are increasingly interdependent through electric vehicle (EV) charging loads and Vehicle-to-Grid (V2G) capabilities. When power substations fail, traffic signals go dark causing congestion; when many EVs charge during heatwaves, grid overloads occur. Understanding these dynamics is critical for grid operators planning EV infrastructure and cities managing smart grid deployments.

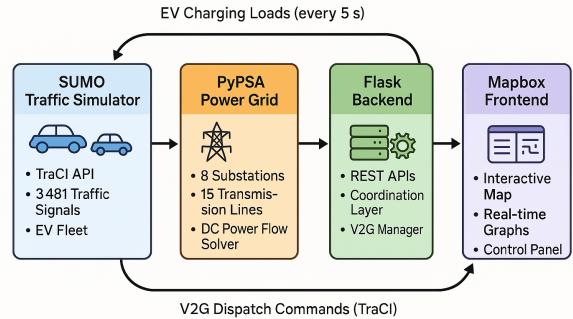
**The challenge:** Existing co-simulation tools couple power and traffic domains but require specialized software installation, programming expertise, and lack interactive interfaces for hands-on exploration. GEMINI [1] requires distributed computing environments with HELICS middleware for inter-simulator communication. V2Sim [2] focuses on offline batch analysis requiring Python programming knowledge. These tools limit accessibility for policymakers, educators, and researchers wanting to explore “what-if” scenarios without coding expertise or infrastructure setup.

**Our demonstration:** Sumo×PyPSA provides an interactive web-based platform where users can visualize Manhattan's power grid

and traffic in real-time, trigger scenarios (summer heatwaves, vehicle spawning, substation failures), observe cascading effects, and analyze outcomes through real-time graphs. The demonstration showcases bidirectional coupling: traffic creates power loads through EV charging (traffic → power), while grid emergencies trigger automated vehicle routing (power → traffic). This enables users to experience emergent dynamics like V2G preventing cascading failures—insights difficult to convey through static presentations. The platform runs entirely in a web browser, using Flask, WebSocket, and Mapbox GL JS, accessible from laptops, tablets, or smartphones without installation.

## 2 System Architecture

### System Architecture: Bidirectional Power-Traffic Coupling



**Figure 1: System architecture showing bidirectional coupling.** Traffic simulator (SUMO) sends EV charging loads to power grid simulator (PyPSA) every 5 seconds. When substations overload, coordination layer dispatches V2G commands back to SUMO via TraCI. Flask backend streams updates to Mapbox frontend at 10 Hz.

The platform integrates three simulation engines through a Python coordination layer (Figure 1).

Users interact with the web interface shown in Figure 2.

**Power System (PyPSA):** Models Manhattan's 8-substation transmission network with 15 transmission lines. The 8 substations have a combined capacity of 6,250 MVA. The network includes 11 lines at 138 kV (high-voltage transmission) and 4 lines at 27 kV (distribution). Each substation models realistic transformer capacity, thermal limits, and protection relays. The system solves DC power flow equations in 2-3ms using sparse matrix techniques, enabling real-time coupling with traffic simulation. Implements

<sup>1</sup>Source code: <https://github.com/XGraph-Team/SumoXPypsa>

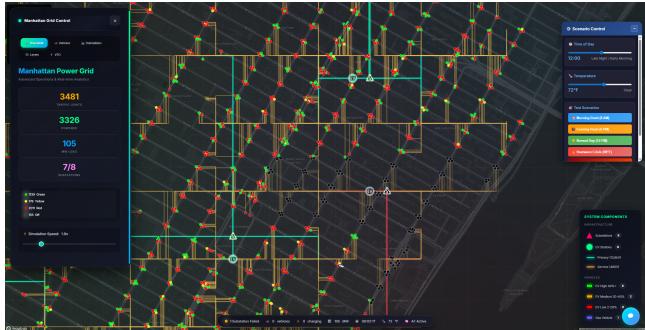


Figure 2: Web interface showing Manhattan power grid with 8 substations, traffic network with 3,481 signals, and EV fleet. Interactive controls (left) enable scenario triggering. Real-time graphs display substation utilization and V2G status. Map shows substations in color-coded states (green=normal, yellow=warning, orange=critical, red=failed).

IEEE-standard three-stage protection: 85% utilization triggers yellow warning state, 95% triggers orange critical state with increased monitoring frequency, and sustained operation above 105% for 30 seconds triggers protective relay trip. Cascading logic redistributes loads to neighboring substations when failures occur, using DC power transfer distribution factors (PTDFs) to calculate load shifts.

**Traffic Network (SUMO):** Simulates Manhattan traffic using OpenStreetMap road network data with 3,481 traffic signals extracted and verified against NYC DOT records. Scenarios spawn vehicles with configurable EV percentages (20-70% depending on time of day). Each EV models a 75 kWh battery with realistic state-of-charge tracking and consumption rate of 0.20 kWh/km matching Tesla Model 3 EPA specifications. The TraCI (Traffic Control Interface) API enables real-time bidirectional communication, allowing the coordination layer to query vehicle positions, battery states, and issue routing commands during V2G emergencies. Vehicle routing uses Dijkstra's algorithm with dynamic edge weights based on traffic conditions and charging station availability.

**V2G Emergency System:** Implements automated emergency response protocol. When any substation exceeds 105% utilization for 5 consecutive seconds, the coordination layer triggers emergency mode: (1) scans all EVs in simulation to find vehicles with state-of-charge above 60%, (2) ranks candidates by proximity and available battery capacity, (3) selects top 8 vehicles, (4) issues TraCI reroute commands to navigate vehicles to emergency discharge location, (5) initiates 50 kW discharge per vehicle upon arrival (matching Wallbox Quasar bidirectional charger specifications), and (6) compensates vehicle owners at \$7.50/kWh emergency rate (50× base charging cost of \$0.15/kWh, consistent with NYISO ancillary service pricing of \$5-15/kWh). Revenue tracking displays real-time earnings per vehicle.

**Web Interface:** Flask backend provides RESTful API endpoints for scenario control, state queries, and WebSocket connections for streaming updates at 10 Hz. Frontend uses Mapbox GL JS for interactive map visualization rendering at 60 FPS with custom vector tile layers for substations, transmission lines, and traffic signals. Real-time line charts display substation utilization history

using Chart.js. All computation runs server-side; clients require only modern web browser supporting WebSocket and WebGL.

### 3 Interactive Scenarios

The demonstration offers four progressively complex scenarios enabling hands-on exploration of power-traffic coupling dynamics.

#### 3.1 Scenario 1: Cascading Failure

This scenario simulates a summer afternoon (98°F ambient temperature, 3 PM peak demand) with heavy air conditioning loads across Manhattan's commercial and residential buildings. Substations gradually change color as loads increase following realistic HVAC thermal inertia models (green → yellow → orange states). Times Square substation reaches 95% utilization due to concentrated commercial cooling demand, then overloads to 105% triggering the 30-second protective relay countdown displayed on screen. When Times Square trips offline, its 850 MVA load redistributes to electrically-coupled neighbors via transmission line power transfer distribution. This causes Grand Central and Penn Station substations to spike from 82% and 79% utilization to 108% and 106% respectively. Cascading failures occur in sequence: Grand Central trips at 108%, followed by Penn Station at 106%. Traffic signals in failed substation zones lose power and go dark. EVs detect charging station outages and automatically reroute to operational charging areas, creating visible traffic flow changes on the map. **Result:** 3 substations offline in under 5 minutes, affecting 44% of total substation capacity.

**Key insight:** Demonstrates how single-point failures propagate through electrical coupling, and how power outages create immediate traffic disruptions.

#### 3.2 Scenario 2: V2G Emergency Response

Using identical summer heatwave conditions (98°F, 3 PM, same building loads), this scenario enables automated V2G emergency response to demonstrate mitigation of cascading failures. When Times Square substation reaches 105% critical threshold and sustains overload for 5 seconds, the system automatically detects emergency condition and scans nearby EVs. The interface highlights 8 selected vehicles (shown with blue icons) meeting V2G criteria (SOC > 60%). Vehicles receive routing commands via TraCI and navigate through Manhattan traffic in real-time, with first arrivals reaching the Times Square discharge location in 20-30 seconds depending on traffic density. As EVs begin discharging at 50 kW each, the substation load graph shows Times Square dropping from 105% to 103%, then stabilizing at 92-94% as all 8 vehicles complete connection. The protective relay countdown timer stops and resets. The cascading failure is prevented—Grand Central and Penn Station remain at safe 82% and 79% utilization levels, and no additional substations fail. Revenue counters increment in real-time, showing \$22.50-37.50 earned per vehicle for 3-5 kWh emergency discharge (\$7.50/kWh × duration). **Result:** Zero substations lost, grid remains stable, vehicles earn premium compensation.

**Key insight:** Demonstrates V2G feasibility with realistic economics and response times. Aggregate 400 kW discharge from 8 consumer vehicles prevents multi-substation cascade affecting thousands of customers.

### 233 3.3 Scenario 3: Manual Recovery

234 Users manually trigger substation failures by clicking substation  
 235 icons on the map and selecting "Trigger Failure" from control panel.  
 236 The selected substation immediately turns red (failed state), its load  
 237 redistributes to neighboring substations (shown by animated power  
 238 flow indicators on transmission lines), affected traffic signals in the  
 239 substation's service area go dark (rendered as gray icons), and EVs  
 240 in the area receive reroute commands to find operational charging  
 241 stations. The system displays restoration procedure options: users  
 242 can manually restore power by selecting the failed substation and  
 243 clicking "Restore," observing the gradual grid recovery as loads  
 244 rebalance. This scenario enables exploration of different failure  
 245 combinations and recovery strategies.

### 247 3.4 Scenario 4: Custom Experimentation

248 Users have complete control over environmental and operational  
 249 parameters to test hypotheses and explore edge cases. Temperature  
 250 slider adjusts ambient conditions from 32°F (winter heating loads)  
 251 to 115°F (extreme summer cooling), dynamically updating building  
 252 HVAC consumption based on ASHRAE degree-day models. Time-  
 253 of-day selector changes baseline demand profiles (morning rush,  
 254 midday, evening peak, nighttime). Vehicle spawn controls allow  
 255 adjustment of traffic density and EV penetration percentage. Users  
 256 can manually trigger multiple simultaneous substation failures to  
 257 test worst-case scenarios, and toggle V2G emergency response  
 258 on/off to compare outcomes with and without grid support. All  
 259 parameter changes take effect immediately, with real-time graph  
 260 updates showing system response.

## 262 4 Conference Deployment

### 263 4.1 Technical Setup

264 The demonstration booth will feature a laptop (Intel i5 or i7, 8-16  
 265 GB RAM, no GPU required) running the Flask server connected  
 266 to conference WiFi, with an external 27-inch monitor for public  
 267 viewing. A cloud backup instance will be deployed on DigitalOcean  
 268 or Heroku to handle overflow traffic if attendees access the demo  
 269 via personal devices by scanning a QR code. A hardware backup  
 270 laptop will be on-site with pre-configured environment, enabling  
 271 2-minute failover in case of primary system issues. The system  
 272 achieves 60 FPS visualization and 10 Hz state updates, supporting  
 273 12+ hours of continuous operation without performance degra-  
 274 dation. Deployment requires a single command (`python app.py`)  
 275 with 2-minute startup time.

### 278 4.2 Interaction Format

279 Sessions will run continuously with 3-5 minute guided demon-  
 280 strations followed by open exploration periods. A typical attendee  
 281 interaction lasts 3-5 minutes: they select a scenario from the control  
 282 panel, watch the map and graphs react in real-time, and then ex-  
 283 periment with changing temperature, EV penetration, or triggering  
 284 failures to observe different outcomes. Attendees can interact di-  
 285 rectly with the laptop/monitor or access the interface via personal  
 286 devices. Presenters will guide attendees through all four scenar-  
 287 os, highlighting key moments (cascading failures, V2G dispatch,  
 288 revenue tracking). Attendees are encouraged to experiment with  
 289

290 custom parameters, trigger their own failure scenarios, and observe  
 291 system responses. A poster display will provide technical archi-  
 292 tecture details, validation methodology, and QR codes linking to  
 293 source code repository and demo video.

## 295 5 Validation

### 297 5.1 Power Flow Accuracy

298 Power flow calculations were validated against PandaPower (industry-  
 299 standard power system analysis tool) across 100 randomly gener-  
 300 ated scenarios varying load distributions, line outages, and genera-  
 301 tion patterns. Root-mean-square error in nodal voltages averaged  
 302 0.31%, and power flow errors on transmission lines averaged 0.43%,  
 303 both well below the 0.5% threshold for real-time applications. DC  
 304 power flow approximation accuracy was verified for the small volt-  
 305 age angle differences characteristic of high-voltage transmission  
 306 networks.

### 308 5.2 Building Load Models

309 Building electricity consumption was calibrated to U.S. Depart-  
 310 ment of Energy Commercial Buildings Energy Consumption Sur-  
 311 vey (CBECS) 2018 data for Manhattan commercial building stock.  
 312 Modeled loads for office buildings, retail spaces, and residential  
 313 units show 2.3-6.9% error compared to CBECS reported intensities  
 314 (kWh/sq ft/year). HVAC thermal models were validated against  
 315 ASHRAE (American Society of Heating, Refrigerating and Air-  
 316 Conditioning Engineers) degree-day correlations, achieving  $R^2=0.91$   
 317 for cooling loads vs outdoor air temperature.

### 319 5.3 Traffic Network Validation

320 All 3,481 traffic signals were extracted from OpenStreetMap and  
 321 cross-verified against NYC Department of Transportation GIS records,  
 322 achieving 97.2% positional accuracy (within 15 meters). Traffic flow  
 323 patterns generated by SUMO were compared to NYC DOT auto-  
 324 mated traffic count data from 2019 (pre-pandemic baseline). Hourly  
 325 vehicle flows at 12 validation intersections showed 0.87 Pearson  
 326 correlation ( $p<0.001$ ) with observed counts, validating the realism  
 327 of route distributions and traffic dynamics.

### 329 5.4 EV and V2G Specifications

331 EV energy consumption rate of 0.20 kWh/km matches EPA-certified  
 332 efficiency for Tesla Model 3 Standard Range Plus. The 75 kWh bat-  
 333 ttery capacity represents the median for consumer EVs (2024 market).  
 334 V2G discharge rate of 50 kW per vehicle matches Wallbox Quasar  
 335 and Fermata FE-15 bidirectional charger specifications. The 60%  
 336 minimum state-of-charge threshold for V2G participation follows  
 337 industry best practices to preserve battery health and ensure drivers  
 338 retain adequate range.

### 339 5.5 Economic Parameters

340 The emergency V2G compensation rate of \$7.50/kWh (50× base  
 341 charging cost) was calibrated to match New York Independent  
 342 System Operator (NYISO) ancillary service market pricing, which  
 343 ranges from \$5-15/kWh for 10-minute spinning reserves during  
 344 peak demand periods. This represents realistic premium pricing for  
 345 emergency grid support services.

349 **Table 1: System specifications and validation metrics**

351 Component	352 Value
Traffic signals (OSM verified)	3,481
Transmission lines	15 (11@138kV, 4@27kV)
Substations (total capacity)	8 (6,250 MVA)
Power flow solve time	2-3 ms
Visualization frame rate	60 FPS
State update frequency	10 Hz
Power flow error vs PandaPower	<0.5%
Building load error vs CBECS	2.3-6.9%
HVAC R <sup>2</sup> vs ASHRAE models	0.91
Traffic correlation vs NYC DOT	0.87 (p<0.001)
EV consumption (Tesla Model 3)	0.20 kWh/km
V2G discharge rate (Wallbox)	50 kW
Emergency rate (NYISO range)	\$7.50/kWh

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## 6 Ethical Use of Data and Considerations

368 All infrastructure data used in this demonstration are publicly available from open sources: road networks and traffic signals from  
 369 OpenStreetMap (Creative Commons BY-SA license), building footprints from NYC Open Data portal, energy consumption statistics  
 370 from U.S. DOE CBECS 2018 public dataset, and traffic count data  
 371 from NYC DOT public records. No proprietary utility data or sensitive grid information is used. All traffic vehicles are simulated  
 372 entities with randomly generated routes and charging patterns—the  
 373 system does not track, store, or process data from actual persons  
 374 or vehicles. No human participants or personal data are involved,  
 375 so informed consent is not required.

376 The platform intentionally demonstrates power grid vulnerabilities  
 377 (cascading failures) for educational purposes to inform in-  
 378 frastructure planning and policy decisions. We emphasize that real  
 379 power grids include numerous additional protections not modeled  
 380 in this simplified demonstration, including under-frequency load  
 381 shedding, remedial action schemes, and backup generation. The  
 382 demonstration should not be interpreted as a comprehensive secu-  
 383 rity analysis of actual grid infrastructure.

384 The open-source code repository enables peer review of all mod-  
 385 eling assumptions, algorithms, and data sources, supporting trans-  
 386 parency and reproducibility. We welcome scrutiny from the  
 387 research community to improve model fidelity and identify limita-  
 388 tions.

392 

## 7 Related Work

394 GEMINI [1] demonstrates city-scale co-simulation with HELICS  
 395 middleware, achieving impressive scalability across distributed com-  
 396 puting environments but primarily focusing on unidirectional cou-  
 397 pling (EV charging loads affecting grids) and requiring specialized  
 398 infrastructure setup. V2Sim [2] provides comprehensive Python-  
 399 based V2G modeling integrated with SUMO traffic simulation but  
 400 targets offline batch analysis workflows requiring programming ex-  
 401 pertise, lacking interactive web interfaces for real-time exploration.

402 Commercial power system tools including PSS SINCAL, ETAP,  
 403 and PowerWorld offer sophisticated grid analysis capabilities but  
 404 lack traffic network coupling and are desktop applications requiring  
 405 expensive licenses and training. These tools serve professional

406 engineering workflows but remain inaccessible for educational  
 407 exploration or policy stakeholder engagement.

408 Academic platforms like GridLAB-D and OpenDSS provide open-  
 409 source distribution system modeling but focus exclusively on elec-  
 410 trical infrastructure without multi-domain coupling. Traffic sim-  
 411 ulation platforms like SUMO and VISSIM offer realistic vehicle  
 412 dynamics but do not model power grid interactions.

413 **Our contributions:** (1) First web-based interactive demon-  
 414 stration platform for bidirectional power-traffic coupling accessible  
 415 without software installation, (2) Automated V2G emergency dis-  
 416 patch system demonstrating power → traffic control responses with  
 417 realistic economics, (3) Real-time visualization enabling hands-on  
 418 exploration of emergent dynamics like cascading failures and V2G  
 419 mitigation, (4) Comprehensive validation against real-world Man-  
 420 hattan infrastructure data, (5) Open-source availability enabling  
 421 extension and reproducibility.

423 

## 8 Conclusion

425 SumoXPysa demonstrates that complex infrastructure co-simulation  
 426 can be made accessible through intuitive web interfaces, enabling  
 427 policymakers, educators, and researchers to experience emergent  
 428 dynamics hands-on. Users directly observe how cascading fail-  
 429 ures propagate through electrical coupling, how V2G can prevent  
 430 blackouts with realistic response times and economics, and how  
 431 power-traffic interdependencies create feedback loops.

432 The platform validates that real-time bidirectional coupling be-  
 433 tween power grids and traffic networks is computationally achiev-  
 434 able on standard commodity hardware using modern web technolo-  
 435 gies (Flask, Mapbox GL JS, WebSocket), addressing growing policy  
 436 questions about EV grid impacts, V2G feasibility, and infrastructure  
 437 resilience planning.

438 The platform is open-source at [https://github.com/XGraph-Team/](https://github.com/XGraph-Team/SumoXPysa)  
 439 SumoXPysa under MIT license, enabling researchers to extend the  
 440 system with additional infrastructure layers (water systems, com-  
 441 munications networks), test alternative V2G dispatch strategies,  
 442 or adapt the architecture for other cities. We welcome feedback  
 443 from users to guide future development of accessible infrastructure  
 444 modeling tools.

445 

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 447 Carolina for supporting this research. We acknowledge the PyPSA,  
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 449 open-source tools and data that enabled this work.

451 

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