

Georouting: An AI-friendly Python Library for Routing and Origin–destination (OD) Matrix Calculation

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Software

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Summary

georouting is a Python package that provides a unified interface to multiple web routing services and open-source local routing solutions, currently including Google Maps, Bing Maps, Esri services, Baidu Maps, TomTom Maps, Here Maps, Mapbox, OpenRouteService, an OSRM local/demo server (Luxen & Vetter, 2011) with helper functions to auto-build and start Docker or Singularity backends (suitable for HPC environments), and routing through OSMnx (Boeing, 2025) and iGraph (Csardi & Nepusz, 2006). It enables researchers and practitioners to compute travel routes, travel times, and travel distances between large numbers of origin–destination (OD) pairs using a consistent and high-level API. The package is designed to integrate naturally with the scientific Python ecosystem by returning results as pandas DataFrames (McKinney & others, 2011) and geopandas GeoDataFrames (Jordahl et al., 2020), and by exposing convenient methods for map-based visualization of routes and accessibility patterns. In many applied domains—such as transport planning, health-care accessibility, logistics, and urban analytics—accurate travel-time and distance estimates are needed to replace simple Euclidean distances or crude network approximations. georouting lowers the barrier to using production-grade web routing services by handling provider-specific request formats, batching strategies, and response parsing behind a single API. For OSRM, it includes Geofabrik region utilities (with optional size lookup) and a one-shot helper that downloads extracts, builds, and launches a local service with the chosen profile (car/foot/bicycle) via Docker or Singularity/Aptainer. The package is available on both PyPI and conda-forge, comes with automated tests and continuous integration, and includes AI-oriented documentation (via llms.txt (Jeremy Howard, 2024)) to make it easier for users to get support from large language models. The package is open source (MIT license) and available at <https://github.com/wybert/georouting>.

Statement of Need

There is a growing need in the health-care, geospatial, and data-science communities for tools that can compute realistic travel distances and travel times at scale. There are many tools available including web-based routing services like Google Maps, Bing/Azure Maps, Esri, Baidu, TomTom, Here, Mapbox, OpenRouteService, and open source routing engines like OSRM and pgRouting (pgRouting contributors, 2025). Researchers often need to compare results from different providers or switch providers due to coverage, accuracy, pricing, or licensing constraints. However, each tool has its own workflow and often limited built-in visualization support, making it difficult to cross verify results or switch providers without significant refactoring. georouting addresses this gap by providing a single, extensible, Pythonic, and AI-friendly routing interface that abstracts over tool-specific details while still exposing tool-specific strengths. Users can change routing methods by switching a class (e.g., GoogleRouter, OSRMRouter, EsriRouter, MapboxRouter, TomTomRouter, HereRouter,

OpenRouteServiceRouter, or an OSMnx-based router) without refactoring their analysis pipeline. Common operations—such as computing a route between two coordinates, building a full OD distance matrix, or computing distances only for specific OD pairs—are implemented in a backend-agnostic way. It also provides a `plot_route()` method to visualize the routing results on an interactive map. We provide `llms.txt` files to make it easier for users to get AI-powered help when working with the package using vibe coding tools like Cursor or Claude Code. For users who need a local engine, georouting ships helper functions to fetch Geofabrik extracts (with optional size metadata), generate an OSRM Dockerfile, and auto-build/run an OSRM backend with the desired profile (car/foot/bicycle) and region via Docker or Singularity/Apptainer (suited to HPC environments). The package is intended for geospatial researchers, transportation and health geographers, operations researchers, and data scientists who need reproducible and scalable routing and accessibility analyses from Python. georouting has been used in the rapidroute project (Xiaokang Fu and Devika Kakkar, 2024), enabling researchers to compute drive times for millions of OD pairs at the national scale within minutes on high-performance computing systems, as well as in a study evaluating drive-time estimation methods on geospatial big data (Fu et al., 2023).

Georouting Functionality and Design

The core of georouting is a set of router classes that inherit from a common `WebRouter` base class. Each router encapsulates the logic for interacting with a specific routing service while sharing a common public interface:

- `get_route(origin, destination)` returns a route object with methods to access total distance, travel time, detailed step-by-step information, and `plot_route` function to visualizing the route on a map, as well as a `GeoDataFrame` representation of the route.
- `get_distance_matrix(origins, destinations, append_od=True)` builds a full distance or duration matrix between sets of origins and destinations, returning a `DataFrame` where rows correspond to OD pairs.
- `get_distances_batch(origins, destinations, append_od=True)` computes distances and durations only for specific OD pairs, using internal batching strategies to respect provider API limits.

The underlying route objects (e.g., `GoogleRoute`, `BingRoute`, `OSRMRoute`) are wrappers around the raw JSON responses returned by each provider. They expose a small, consistent set of methods such as `get_distance()`, `get_duration()`, `get_route()`, and `get_route_geopandas()`. For map-based visualization, route objects can be plotted directly, producing interactive maps (e.g., via `folium`) suitable for exploratory analysis in a Jupyter environment (see Figure 1).

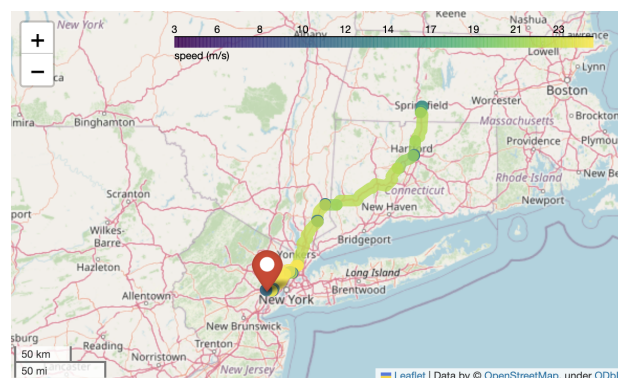


Figure 1: Figure.

To support large OD problems, georouting implements helper utilities that partition OD pairs

78 into batches that respect provider-specific API limits while avoiding redundant calls. These
 79 utilities work with vectorized data structures (e.g., Numpy arrays or pandas DataFrames)
 80 and are designed to minimize the number of remote API requests. For users needing on-
 81 prem/HPC execution, georouting ships OSRM helpers that can download Geofabrik extracts,
 82 build images, and launch local OSRM backends via Docker or Singularity/Apptainer—allowing
 83 high-throughput OD computation without relying on external APIs. The package is written in
 84 pure Python and integrates closely with widely used geospatial libraries such as geopandas and
 85 osmnx, enabling workflows that combine web-based routing with open street network analysis.

86 The project follows standard Python packaging practices, ships with an automated test suite
 87 that covers core routing and batching logic (including OSRM-based tests that do not require
 88 API keys), and is continuously tested on multiple Python versions using GitHub Actions.
 89 Documentation is built with MkDocs, including narrative tutorials, usage examples, and API
 90 reference pages generated from docstrings. In addition, the project maintains `llms.txt` and
 91 `llms-full.txt` files to provide machine-readable documentation for large language models,
 92 making it easier for users to obtain high-quality automated assistance.

93 Acknowledgements

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 95 2025) for geocoding in Python, as well as by the growing ecosystem of open-source geospatial
 96 tools such as geopandas and osmnx. The author thanks the maintainers of these projects.
 97 The author also thanks early users and contributors who provided feedback on API design,
 98 documentation, and test coverage.

99 References

- 100 Boeing, G. (2025). Modeling and analyzing urban networks and amenities with OSMnx.
 101 *Geographical Analysis*, 57(4), 567–577. <https://doi.org/10.1111/gean.70009>
- 102 Contributors, G. (2025). *Geopy/geopy*. geopy.
- 103 Csardi, G., & Nepusz, T. (2006). The igraph software package for complex network research.
 104 *InterJournal, Complex Systems*, 1695.
- 105 Fu, X., Kakkar, D., Chen, J., Moynihan, K., Hegland, T., & Blossom, J. (2023). A comparative
 106 study of methods for drive time estimation on geospatial big data: A case study in USA.
 107 *The International Archives of the Photogrammetry, Remote Sensing and Spatial Information*
 108 *Sciences*, 48.
- 109 Jeremy Howard. (2024). *The /llms.txt file*. <https://github.com/answerdotai/llms-txt>
- 110 Jordahl, K., Bossche, J. V. den, Fleischmann, M., Wasserman, J., McBride, J., Gerard,
 111 J., Tratner, J., Perry, M., Badaracco, A. G., Farmer, C., Hjelle, G. A., Snow, A. D.,
 112 Cochran, M., Gillies, S., Culbertson, L., Bartos, M., Eubank, N., maxalbert, Bilogur,
 113 A., ... Leblanc, F. (2020). *Geopandas/geopandas: v0.8.1* (Version v0.8.1). Zenodo.
 114 <https://doi.org/10.5281/zenodo.3946761>
- 115 Luxen, D., & Vetter, C. (2011). Real-time routing with OpenStreetMap data. *Proceedings*
 116 *of the 19th ACM SIGSPATIAL International Conference on Advances in Geographic*
 117 *Information Systems*, 513–516. <https://doi.org/10.1145/2093973.2094062>
- 118 McKinney, W., & others. (2011). Pandas: A foundational python library for data analysis and
 119 statistics. *Python for High Performance and Scientific Computing*, 14(9), 1–9.
- 120 pgRouting contributors. (2025). *pgRouting: Routing on PostgreSQL*. <https://doi.org/10.5281/zenodo.15004469>
- 121

¹²² Xiaokang Fu and Devika Kakkar. (2024). *Rapidroute*. [https://rapidrouteapp.onrender.com/](https://rapidrouteapp.onrender.com/routing_app)
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