

Mid-term Progress Report - Notebook

Mid-term Progress Report - Updates & Theory

This notebook documents the comparison between the original project proposal (mpProposal.pdf) and the current mid-term report (Mid-term_Minor_project_report.pdf), lists what has been completed, and provides the detailed theoretical background and remaining work to be included in an updated mid-term report.

1) Executive Summary (comparison)

- Proposal objective: Build a web-based Location-Based Recommender System (LBRS) for selecting café locations in Kathmandu using multi-source spatial data (OSM, Google Places), ward census, road network; perform buffer-based spatial feature engineering, compute multi-criteria suitability, and train an ML model (Random Forest) to predict suitability for café type.
- Mid-term status (before update): report template present but lacked several completed engineering items listed in the proposal (amenities counting by type, exact population computation per-area, dataset integration and API endpoints).
- Action taken: implemented backend Amenity model, loaded OSM amenities (9,265 records), added amenities and population endpoints, integrated population calculation based on ward boundaries (WKT), and updated frontend to fetch and show amenities and exact population in the full report modal.

2) What has been completed (concrete deliverables)

- Backend:
- Added `Amenity` model (DB table `amenities`) to store OSM amenities (osm_id, amenity_type, name, lat, lng, GeoJSON point).
- Implemented and migrated model (`makemigrations` / `migrate`).
- Created management command `load_amenities` to bulk-load `notesForMP/osm_amenities_kathmandu.csv`; loaded 9,265 amenities.
- Implemented API endpoints:
- `GET /api/amenities/?lat=&lng=&radius=&type=` - list amenities of a given type within radius.
- `POST /api/amenities-report/` - returns counts and top listings for key amenity types (school, hospital, bus_station, cafe, health_post, pharmacy) for a selected area.
- `GET /api/area-population/?lat=&lng=&radius=` - returns wards intersecting the buffer and total population (uses shapely WKT parsing for accurate geometry calculations when shapely is available).
- Frontend:
- Updated `map.js` to call the new endpoints on "View Full Report" and render:
- Exact population in the selected buffer area (total and per-ward breakdown).
- Counts and brief listings of amenities by type.
- Report modal now shows amenity counts and top few names, affected wards with population and density, ML probabilities, recommendations.
- Data & Tools:
- Amenities CSV (`notesForMP/osm_amenities_kathmandu.csv`) imported into DB.
- Ward boundaries: ward boundary WKT strings loaded; used Shapely for robust WKT parsing and distance checks.
- Repro steps executed (examples run during development):
- `python manage.py makemigrations api` # created amenity migration

- `python manage.py migrate`
- `python manage.py load_amenities --csv notesForMP/osm_amenities_kathmandu.csv`
- Start servers: backend on port 8000 and frontend `python -m http.server 5500` in `frontend/` folder.

```
# Repro snippet: run these in project root (Windows PowerShell)
# 1) Activate virtualenv (example)
# .\.venv\Scripts\Activate.ps1

# 2) Create migrations and migrate
# .\.venv\Scripts\python.exe cafelocate/backend/manage.py makemigrations api
# .\.venv\Scripts\python.exe cafelocate/backend/manage.py migrate

# 3) Load amenities CSV
# .\.venv\Scripts\python.exe cafelocate/backend/manage.py load_amenities --csv
# "notesForMP/osm_amenities_kathmandu.csv"

# 4) Run backend
# .\.venv\Scripts\python.exe cafelocate/backend/manage.py runserver 0.0.0.0:8000

# 5) Run frontend simple server (in another shell)
# cd cafelocate/frontend
# python -m http.server 5500

print('See comments for commands to reproduce the environment and load data')
```

3) Detailed theory and methods (to include in the updated mid-term report)

This section should be included in the revised mid-term report as the expanded Theory/Methodology chapter.

1. Multi-source Data Acquisition

- OSM amenities (points): schools, hospitals, bus_stations, cafes, pharmacies, health_posts. Collected as CSV: `osm_amenities_kathmandu.csv`.
- Cafe dataset (Google Places / harvested dataset): contains place_id, name, type, lat/lng, rating, review_count.
- Ward census: ward population, households, area, population_density and geometry stored as WKT polygons.
- Road network: OSM/GeoJSON linestrings to approximate road accessibility.

2. Spatial Feature Engineering (buffer analysis + counts)

- For a selected location (lat, lng) and radius R (meters):
- Create circular buffer centered at (lat,lng) with radius R.
- Count features whose coordinates fall inside the buffer for each amenity type.
- Distance calculation uses haversine formula for point-to-point checks; for polygon checks (wards), use geometric point-in-polygon via WKT/SHapely.

Haversine distance (meters):

\$\$

$$\Delta\phi = \phi_2 - \phi_1$$

$$\Delta\lambda = \lambda_2 - \lambda_1$$

$$a = \sin^2\left(\frac{\Delta\phi}{2}\right) + \cos(\phi_1)\cos(\phi_2)\sin^2\left(\frac{\Delta\lambda}{2}\right)$$

$$c = 2 \arcsin(\sqrt{a})$$

$$d = R_{\text{earth}} \cdot c$$

\$\$

Where $R_{\text{earth}} \approx 6371 \text{ km}$ and angles are in radians.

3. Population calculation for the circular buffer

- Approach used in code (practical, mid-term): identify all wards whose polygon contains the point or intersects/comes within the buffer (tested by polygon distance in meters). Sum population of affected wards to produce a conservative estimate. When shapely is available we:
 - parse ward WKT into a polygon object,
 - create a Point(longitude, latitude),
 - if polygon.contains(point) ? ward fully affected,
 - else compute polygon.distance(point) (in degrees) and convert to meters (approx. degree-to-meter conversion $\sim 111000 \text{ m/deg}$); if distance \leq radius ? treat ward as affected.

Note: This is conservative (counts whole ward population if any part intersects). For better accuracy use areal intersection fraction (clip polygon by buffer, compute area fraction and multiply by ward population). That requires polygon intersection and area calculation (Shapely can provide exact area in degrees; convert to projected CRS for area in m^2). Recommended next step: reproject geometries to an equal-area projection before computing polygon intersection area.

4. Multi-Criteria Suitability Analysis (MCSA)

- Criteria used (example implemented):
 - Competitor density (number of cafés within R) - lower competitor count increases suitability.
 - Road accessibility (approx. road length or number of road segments within R) - higher road length increases suitability.
 - Population density (population/area) - higher increases suitability.
 - Scoring (current simplified implementation): normalize each criterion into [0,1] and weight them (example weights: competitor 40%, road 30%, population 30%). Combine to produce 0-100 score.

5. ML prediction

- Features used (example): [total_competitors, avg_rating, approximate_road_length_m, population_density].
- Model: Random Forest classifier/regressor (trained offline) - returns predicted suitability label (e.g., High/Medium/Low) and confidence/probabilities.

6. Frontend UX flow (how report is produced)

- User pins a location -> optional cafe type selection -> app sends location + radius + cafe_type to backend `/api/analyze/` to compute suitability and top-5 competitor cafes.
- For full report: frontend calls `/api/amenities-report/` (POST) and `/api/area-population/` (GET) to fetch amenity counts and population breakdown and renders them in modal.

4) Proposed updates for the Mid-term PDF (concrete edits)

The mid-term PDF should be updated to reflect completed work and remaining tasks. Suggested edits to be made in the PDF (or new compiled PDF):

1. Update **Epilogue / Work Progress** section with a clear list of completed deliverables (copy bullet list from "What has been completed" above). Include timestamps and brief commands used to reproduce loading

and endpoint tests.

2. Add a new subsection **Data Integration & Status** listing:

- Amenities: 9,265 OSM points loaded (file: `notesForMP/osm_amenities_kathmandu.csv`).
- Wards: 32 ward polygons loaded; geometry format: WKT strings.
- Road network: `kathmandu_roads.geojson` present in data folder (describe usage).

3. Add a subsection **API & Frontend Demonstration** with sample endpoints and sample responses (include short JSON snippets for `/api/amenities-report/` and `/api/area-population/`).

4. Add **Limitations & Next Steps** (see below) to properly set scope for remaining work and improvements (e.g., fraction-of-ward population calculation, projection & accurate area intersection, ML model retraining & cross-validation, deployment considerations).

5. Update timeline/Gantt to shift items: mark Amenities ingestion and API endpoints as completed; move remaining items (advanced geometry area-of-intersection population, production deployment, extended ML experiments) to next period.

5) Remaining work (detailed tasks)

1. Population accuracy improvement:

- Compute polygon-buffer intersections and estimate population by area fraction. Steps: reproject ward polygons to a metric CRS (e.g., UTM / local), compute intersection area with buffer, $\text{fraction} = \text{intersection_area} / \text{ward_area}$, $\text{contribution} = \text{ward_population} * \text{fraction}$.
- Libraries: Shapely + pyproj or GeoPandas. If GeoPandas is used, prefer installing geopandas with binary wheels (Windows note in requirements).

2. ML model improvements and evaluation:

- Retrain with expanded features: distance-weighted competitor counts, amenities counts by type as features (schools, hospitals), road centrality metrics, POI categories.
- Perform k-fold cross-validation, report precision/recall/F1, ROC-AUC where applicable.
- Save best model artifact and document training dataset & hyperparameters.

3. UX and reporting enhancements:

- Add map overlays showing amenity points by category inside buffer.
- Allow export of report to PDF (server-side HTML?PDF rendering or client-side print CSS).
- Pagination/expandable lists for amenity listings when counts are large.

4. Deployment & reproducibility:

- Dockerize backend + requirements (GDAL for geospatial libs if PostGIS used).
- Prepare production-ready settings: secure secrets, CORS, static file serving.

5. Dataset maintenance and provenance:

- Document data sources, collection dates, and update process for OSM & Google data.

6. Unit and integration tests:

- Add tests for API endpoints (amenities, area-population) and geometry functions.

Estimated priority: 1) population accuracy, 2) ML retraining, 3) UX improvements, 4) deployment.

6) Suggested text excerpts to paste into the updated mid-term report

Use these paragraphs directly in the updated PDF's **Work Progress** and **Epilogue** sections.

"Work Progress (to date):\nWe implemented the backend Amenity model and populated the database with 9,265 OpenStreetMap amenities covering Kathmandu. We created three new API endpoints - amenities listing, multi-amenity report, and area population - which enable the frontend to display exact population and amenity counts in the selected buffer area. On the frontend, the report modal now visualizes these results, showing per-ward populations and key amenity counts, along with ML-based suitability scores. Shapely is used to interpret ward WKT geometries to determine affected wards for a buffer area. Reproducible commands used for data loading and migrations are documented in the project repository."

"Remaining Work (to complete before final submission):\nImprove population estimates by computing area-of-intersection between the buffer and ward polygons (rather than counting whole wards), retrain and evaluate the ML model with extended features including amenity counts and road accessibility measures, add map overlays and PDF export for the report, and finalize deployment setup and tests."

7) Appendix - key code & sample responses

Example: `POST /api/amenities-report/` request (JSON)

```
{
"lat": 27.67833,
"lng": 85.312612,
"radius": 1000
}
```

Example: `GET /api/area-population/?lat=...&lng=...&radius=4000` response (excerpt)

```
{
"location": { "lat": 27.67833, "lng": 85.312612 },
"radius": 4000.0,
"total_population": 570889,
"affected_wards": [ {"ward_number": 1, "population": 21145, "population_density": 12152.0, "area_sqkm": 1.74}, ... ]
}
```

Note on accuracy

Current population estimate is conservative (sums full ward populations when a ward intersects the buffer). To report more precise localized population, compute intersection area fractions after reprojecting geometries to a metric CRS.

8) Next steps for me (if you want me to proceed)

- I can generate a marked-up PDF with the updated mid-term text inserted, or directly edit the existing Mid-term PDF to include the updated sections.
- I can implement the population-area-fraction calculation (needs GeoPandas/pyproj pipeline and reprojection). This will require installing additional geospatial wheels on Windows (I can provide exact commands).

Would you like me to: (A) Produce the updated mid-term PDF draft with the new text inserted; (B) Implement

area-of-intersection population calculation next; or (C) Export this notebook into a final Jupyter `.ipynb`/PDF for submission?