

Earthquake Prediction System Documentation

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A comprehensive guide to the Earthquake Prediction System, a web-based application for predicting earthquake metrics and visualizing data.

For support, visit: <https://x.ai>

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1 Overview

The Earthquake Prediction System is a web-based application built using Django that allows users to input earthquake data (latitude, longitude, magnitude, and depth) to generate predictions. The system provides a prediction dashboard, a data visualization interface, a real-world map display, and the ability to download prediction data as a PDF report. It leverages machine learning for magnitude prediction, geocoding for location names, and interactive visualizations using Chart.js and Leaflet.

- **Developed by:** CHOGE VINCENT
- **Date:** May 04, 2025
- **Technologies:** Python, Django, HTML, JavaScript, Chart.js, Leaflet, reportlab, NumPy, scikit-learn, geopy

2 System Architecture

The application follows a Model-View-Controller (MVC) pattern, implemented via Django's framework:

- **Models:** Define the data structure (**Earthquake** and **Prediction** models) stored in a SQLite database by default.
- **Views:** Handle HTTP requests and responses, including data processing, prediction generation, and PDF creation.
- **Templates:** Provide the HTML/CSS/JavaScript frontend for user interaction.
- **Static Files:** Include stylesheets (Bootstrap, custom CSS) and JavaScript libraries (Chart.js, Leaflet).

2.1 Key Components

2.1.1 Database Models

- **Earthquake:** Stores raw input data (latitude, longitude, magnitude, depth, date).
- **Prediction:** Stores predicted values (predicted magnitude, confidence, probability, accuracy, predicted location, prediction date) linked to an **Earthquake** instance.

2.1.2 Views

- **home:** Renders the homepage.
- **dashboard:** Displays earthquake inputs and predictions in tables with a map.
- **predict:** Handles form submission, generates predictions, and displays results with graphs and maps.
- **download_pdf:** Generates a landscape PDF with a table of all predictions.
- **admin_dashboard:** Provides an admin interface to delete records.
- **delete_earthquake/delete_prediction:** Handles deletion of records.

2.1.3 Templates

- **home.html:** Homepage with navigation.
- **dashboard.html:** Shows tables and a map of all data.
- **predict.html:** Form for input, prediction display, bar graphs, and map.

- `admin.html`: Admin interface for record management.

2.1.4 External Libraries

- **Chart.js**: For rendering bar graphs of prediction metrics.
- **Leaflet**: For displaying real-world maps with markers.
- **reportlab**: For generating PDF reports.
- **geopy**: For geocoding coordinates to location names.
- **NumPy/scikit-learn**: For simple linear regression-based magnitude prediction.

3 How It Works

3.1 Data Input and Prediction Generation

Process:

1. Users access the “Make Prediction” page (`/predict/`) and enter earthquake data:
 - Latitude (with direction: N/S)
 - Longitude (with direction: E/W)
 - Magnitude
 - Depth (km)
2. The `predict` view validates the input (e.g., latitude between -90 and 90).
3. It checks for duplicates within the last 30 days.
4. The data is saved as an **Earthquake** instance.
5. A predicted magnitude is calculated using a simple linear regression model based on historical data (if available; otherwise, it uses the input magnitude).
6. Confidence (80–95%) and probability (70–90%) are randomly generated as placeholders (to be replaced with a real model).
7. Accuracy is computed as the average of confidence and probability.
8. The predicted location is determined via geopy’s Nominatim service (with a fallback to coordinates if geocoding fails).
9. A **Prediction** instance is saved with all calculated values.

Output:

- The exact predicted location is displayed.
- Small horizontal bar graphs show magnitude, confidence, probability, and accuracy.
- A Leaflet map displays the location with a marker.

3.2 Data Visualization

Dashboard (`/dashboard/`):

- Displays two tables:
 - “Recorded Earthquake Inputs”: Lists all **Earthquake** data.
 - “Prediction Results”: Lists all **Prediction** data, including accuracy.
- A Leaflet map shows markers for all predicted locations.

Graphs:

- Generated using Chart.js, showing metrics as horizontal bars for clarity.

3.3 PDF Report Generation

Process:

1. Users access the “Download & Print PDF” link (`/download_pdf/`).
2. The `download_pdf` view retrieves all `Prediction` instances.
3. A landscape PDF is created using `reportlab`, containing:
 - A title with the generation date.
 - A table with columns: Location, Magnitude, Confidence (%), Probability (%), Accuracy (%).
4. The PDF is sent as a downloadable file (“`predictions.pdf`”).

Features:

- Landscape orientation for wider table display.
- Text wrapping in the Location column to handle long names.
- Styled table with grid lines, centered text, and color coding.

3.4 Administration

Admin Dashboard (`/admin_dashboard/`):

- Allows deletion of `Earthquake` and `Prediction` records via POST requests.
- Displays all records in tables for manual review.

4 Setup Instructions

4.1 Prerequisites

- Python 3.13.2 (or compatible version)
- pip (Python package manager)
- Git (optional, for version control)

4.2 Installation

1. **Clone the Repository** (if applicable):

```
git clone <repository-url>
cd earthquake_project
```
2. **Create a Virtual Environment:**

```
C:\Users\Butler\Desktop\EARTHQUAKE\earthquake_project\env\Scripts\activate
```
3. **Install Dependencies:**

```
pip install django==5.2 numpy scikit-learn geopy reportlab
```
4. **Apply Migrations:**

```
python manage.py makemigrations
python manage.py migrate
```
5. **Run the Server:**

```
python manage.py runserver
```

Access the app at <http://127.0.0.1:8000/>.

4.3 Configuration

- **Database:** Uses SQLite by default (configured in `settings.py`). For production, modify `DATABASES` to use PostgreSQL or MySQL.
- **Static Files:** Ensure `STATIC_URL` and `STATICFILES_DIRS` are set in `settings.py`.
- **Logging:** Configured in `views.py` to log errors and info to the console.

5 Usage Guidelines

5.1 User Workflow

1. **Navigate to Home:**
 - Visit <http://127.0.0.1:8000/> for an overview and navigation links.
2. **Make a Prediction: Symptom**
 - Go to <http://127.0.0.1:8000/predict/>.
 - Enter earthquake data and click “Predict”.
 - Review the location, graphs, and map.
3. **View Dashboard:**
 - Go to <http://127.0.0.1:8000/dashboard/>.
 - Check tables and map for all records.
4. **Download PDF:**
 - Go to http://127.0.0.1:8000/download_pdf/.
 - Download and print the landscape PDF with all prediction data.
5. **Admin Actions** (if authorized):
 - Go to http://127.0.0.1:8000/admin_dashboard/.
 - Delete records as needed.

5.2 Developer Notes

- **Extending Predictions:** Replace the random confidence/probability with a machine learning model (e.g., Random Forest) for better accuracy.
- **Geocoding:** Handle rate limits or failures by caching results or using a paid API (e.g., Google Maps).
- **Scalability:** For large datasets, optimize database queries or switch to a NoSQL database.
- **Testing:** Add unit tests for views and models using Django’s testing framework.

6 Troubleshooting

- **PDF Not Downloading:**
 - Ensure `reportlab` is installed (`pip install reportlab`).
 - Check logs for errors in `views.py`.
- **Graphs/Map Not Loading:**
 - Verify internet access for CDN libraries (Chart.js, Leaflet).
 - Check browser console for JavaScript errors.

- **Database Errors:**

- Run `python manage.py migrate` to apply schema changes.
- Clear the database with `python manage.py flush` if needed.

7 Future Enhancements

- **Real-Time Data:** Integrate with seismic data APIs (e.g., USGS).
- **Advanced Models:** Use deep learning for more accurate predictions.
- **User Accounts:** Add authentication for personalized data tracking.
- **Mobile App:** Develop a companion app using Django REST Framework.

8 Contact

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