



Learning the SIR Epidemic Model

Python

3.10+



PyTorch



FastAPI



Tailwind CSS

GSoC 2026 Proposal Prototype for Human AI Foundation > *Project:* Deducing deterministic ODEs of the SIR model from stochastic simulations using Machine Learning and Auto-differentiation.



Project Overview

The classic Susceptible-Infected-Removed (SIR) epidemic model is defined by a set of Ordinary Differential Equations (ODEs). However, real-world epidemics are often stochastic. This project aims to bridge that gap by using **Machine Learning** to deduce the deterministic form of the SIR model from a large number of synthetic, stochastically generated epidemics.

This repository contains the prototype built for the GSoC '26 proposal, demonstrating the end-to-end pipeline:

1. Generating stochastic SIR data (Gillespie/Binomial approximation).
2. Processing data through a PyTorch-based Neural Network.
3. Using Auto-Differentiation to extract $\frac{dS}{dt}$, $\frac{dI}{dt}$, and $\frac{dR}{dt}$.
4. Visualizing the comparison via a modern Web UI.



Tech Stack

- **Backend:** Python, FastAPI, Uvicorn
- **Machine Learning & Math:** PyTorch (for Auto-diff & PINNs), NumPy
- **Frontend:** HTML5, TailwindCSS, JavaScript
- **Data Visualization:** Chart.js



Project Structure

```
gsoc-sir-project/
├── backend/
│   ├── main.py           # FastAPI server & API endpoints
│   ├── simulator.py      # Stochastic SIR simulation logic
│   ├── ml_model.py       # PyTorch Neural Network & Auto-diff logic
│   └── requirements.txt  # Python dependencies
└── frontend/
    └── index.html        # Interactive Dashboard UI
```



Installation & Setup

Follow these steps to run the project on your local machine.

1. Clone the repository

```
git clone https://github.com/PritamHazra2708/SIR-Model-For-HumanAI.git
```

2. Setup the Backend

Navigate to the backend directory and install the required dependencies:

```
cd backend
python -m venv venv
source venv/bin/activate # On Windows use: venv\Scripts\activate
pip install fastapi uvicorn torch numpy pydantic
```

3. Run the FastAPI Server

Start the backend server using Uvicorn:

```
uvicorn main:app --reload
```

The API will be available at: <http://127.0.0.1:8000>

4. Run the Frontend

Simply open the [frontend/index.html](#) file in any modern web browser. No frontend build tools are required for this prototype.

Core Highlight: Auto-Differentiation

A major requirement of this project is using symbolic methods to approximate $S(t)$, $I(t)$, $R(t)$. In [backend/ml_model.py](#), we utilize PyTorch's [autograd](#) engine to compute derivatives directly from the neural network's graph:

```
# Extracting exact derivatives learned by the model
dS_dt = torch.autograd.grad(S_pred, inputs, create_graph=True)
dI_dt = torch.autograd.grad(I_pred, inputs, create_graph=True)
dR_dt = torch.autograd.grad(R_pred, inputs, create_graph=True)
```

Contributing

This is an open-source initiative under the Human AI Foundation. Suggestions and pull requests are welcome!

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