

Wildfire Analysis: End-to-End Blueprint

✓ Phase 1: Core Pipeline (COMPLETED)

What We've Built:

1. Data Extraction Module (`(data_extraction.py)`)

- Unified interface for VIIRS, ERA5, and DEM data
- Spatial clipping and coordinate transformation
- Feature extraction with error handling

2. Configuration System (`(config_module.py)`)

- Centralized configuration management
- Easy customization of parameters
- Environment-based API key loading

3. Transformer Model (`(transformer_model.py)`)

- PyTorch-based transformer architecture
- Training pipeline with validation
- Model checkpointing and loading

4. Main Pipeline (`(main_pipeline.py)`)

- Command-line interface
- Step-by-step or full pipeline execution
- Support for both transformer and Random Forest

5. Testing Suite (`(test_script.py)`)

- Comprehensive tests for all components
- Synthetic data testing
- End-to-end validation

🚀 Phase 2: Local Testing (NEXT STEP)

Tasks:

1. Setup Environment

```
bash

# Create virtual environment
python -m venv wildfire-env
source wildfire-env/bin/activate # On Windows: wildfire-env\Scripts\activate

# Install dependencies
pip install -r requirements.txt

# Verify installation
python test_script.py
```

2. Run Pipeline with Your Data

```
bash

# Full pipeline with transformer
python main_pipeline.py --step full --model transformer

# Check outputs
ls outputs/
# Should see:
# - wildfire_features.csv
# - transformer_model.pth
# - wildfire_predictions.csv
```

3. Evaluate Results

```
python
```

```

import pandas as pd

# Load predictions
df = pd.read_csv('outputs/wildfire_predictions.csv')

# Check accuracy
accuracy = (df['confidence_num'] == df['predicted_confidence']).mean()
print(f"Accuracy: {accuracy:.2%}")

# Confusion matrix
confusion = pd.crosstab(
    df['confidence'],
    df['predicted_confidence_label']
)
print(confusion)

```

4. Experiment with Hyperparameters

```

python

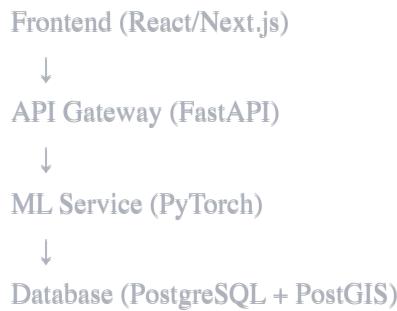
# In config_module.py, try:
model.d_model = 256 # Increase model capacity
model.num_layers = 6 # Deeper network
model.dropout = 0.2 # More regularization

# Then retrain
python main_pipeline.py --step train --model transformer

```

Phase 3: Open-Source Website

Architecture Overview



3.1 Backend API (FastAPI)

File: [api/main.py](#)

python

```
from fastapi import FastAPI, File, UploadFile
from fastapi.middleware.cors import CORSMiddleware
import torch
import pandas as pd
from transformer_model import WildfireTransformer, WildfireModelTrainer

app = FastAPI(title="Wildfire Prediction API")

# CORS for frontend
app.add_middleware(
    CORSMiddleware,
    allow_origins=["*"],
    allow_methods=["*"],
)

# Load model at startup
@app.on_event("startup")
async def load_model():
    global model, trainer
    model = WildfireTransformer(input_dim=5)
    trainer = WildfireModelTrainer(model)
    trainer.load_checkpoint('outputs/transformer_model.pth')

# Health check
@app.get("/")
async def root():
    return {"status": "online", "model": "wildfire_transformer"}

# Single prediction
@app.post("/predict")
async def predict(
    frp: float,
    u10: float,
    v10: float,
    elevation: float,
    slope: float
):
    features = [[frp, u10, v10, elevation, slope]]
    prediction = trainer.predict(features)[0]
    confidence_map = {0: "low", 1: "nominal", 2: "high"}


    return {
        "prediction": confidence_map[prediction],
```

```

"confidence_score": int(prediction),
"input_features": {
    "frp": frp,
    "wind_u": u10,
    "wind_v": v10,
    "elevation": elevation,
    "slope": slope
}
}

# Batch prediction
@app.post("/predict_batch")
async def predict_batch(file: UploadFile = File(...)):
    df = pd.read_csv(file.file)
    features = df[['frp', 'u10', 'v10', 'elevation', 'slope']].values
    predictions = trainer.predict(features)

    df['predicted_confidence'] = predictions
    return df.to_dict(orient='records')

# Model statistics
@app.get("/stats")
async def get_stats():
    return {
        "training_history": trainer.history,
        "model_parameters": {
            "d_model": 128,
            "num_layers": 4,
            "nhead": 8
        }
    }

```

Run API:

```

bash
uvicorn api.main:app --reload --port 8000

```

3.2 Frontend (React/Next.js)

File: [frontend/pages/index.js](#)

```

jsx

```

```
import { useState } from 'react';
import dynamic from 'next/dynamic';

// Lazy load map component
const MapComponent = dynamic(() => import('../components/Map'), {
  ssr: false
});

export default function Home() {
  const [prediction, setPrediction] = useState(null);
  const [loading, setLoading] = useState(false);

  const [formData, setFormData] = useState({
    frp: 10,
    u10: 2.5,
    v10: 3.0,
    elevation: 500,
    slope: 30
  });

  const handlePredict = async () => {
    setLoading(true);
    try {
      const response = await fetch('http://localhost:8000/predict', {
        method: 'POST',
        headers: { 'Content-Type': 'application/json' },
        body: JSON.stringify(formData)
      });
      const data = await response.json();
      setPrediction(data);
    } catch (error) {
      console.error('Prediction failed:', error);
    } finally {
      setLoading(false);
    }
  };
}

return (
  <div className="container">
    <h1>Wildfire Confidence Predictor</h1>

    <div className="input-panel">
      <label>
```

```

Fire Radiative Power (MW):
<input
  type="number"
  value={formData.frp}
  onChange={(e) => setFormData({ ...formData, frp: e.target.value})}
/>
</label>

<label>
Wind U Component (m/s):
<input
  type="number"
  value={formData.u10}
  onChange={(e) => setFormData({ ...formData, u10: e.target.value})}
/>
</label>

{/* More inputs... */}

<button onClick={handlePredict} disabled={loading}>
  {loading ? 'Predicting...' : 'Predict Confidence'}
</button>
</div>

{prediction && (
  <div className={`result ${prediction.prediction}`}>
    <h2>Prediction: {prediction.prediction.toUpperCase()}</h2>
    <p>Confidence Score: {prediction.confidence_score}/2</p>
  </div>
)}

<MapComponent predictions={prediction} />
</div>
);
}

```

3.3 Interactive Map Component

File: [frontend/components/Map.js](#)

jsx

```
import { MapContainer, TileLayer, Marker, Popup, CircleMarker } from 'react-leaflet';
import 'leaflet/dist/leaflet.css';

export default function Map({ fireData }) {
  const center = [34.05, -118.25];

  const getColor = (confidence) => {
    return {
      low: '#808080',
      nominal: '#FFA500',
      high: '#FF0000'
    }[confidence];
  };

  return (
    <MapContainer
      center={center}
      zoom={9}
      style={{ height: '500px', width: '100%' }}
    >
    <TileLayer
      url="https://s.tile.openstreetmap.org/{z}/{x}/{y}.png"
      attribution='&copy; OpenStreetMap contributors'
    />

    {fireData && fireData.map((fire, idx) => (
      <CircleMarker
        key={idx}
        center={[fire.latitude, fire.longitude]}
        radius={fire.frp / 5}
        fillColor={getColor(fire.predicted_confidence)}
        color={getColor(fire.predicted_confidence)}
        fillOpacity={0.7}
      >
      <Popup>
        <strong>FRP:</strong> {fire.frp} MW<br/>
        <strong>Confidence:</strong> {fire.predicted_confidence}<br/>
        <strong>Elevation:</strong> {fire.elevation}m<br/>
        <strong>Slope:</strong> {fire.slope}°
      </Popup>
    </CircleMarker>
  )))
  </MapContainer>
}
```

```
 );  
 }
```

3.4 Database Schema (PostgreSQL + PostGIS)

```
sql  
  
-- Fire detections table  
CREATE TABLE fire_detections (  
    id SERIAL PRIMARY KEY,  
    latitude DOUBLE PRECISION NOT NULL,  
    longitude DOUBLE PRECISION NOT NULL,  
    geometry GEOMETRY(Point, 4326),  
    frp REAL,  
    confidence TEXT,  
    predicted_confidence TEXT,  
    u10 REAL,  
    v10 REAL,  
    elevation REAL,  
    slope REAL,  
    detection_time TIMESTAMP,  
    created_at TIMESTAMP DEFAULT NOW()  
);  
  
-- Spatial index  
CREATE INDEX idx_fire_geom ON fire_detections USING GIST (geometry);  
  
-- Predictions history  
CREATE TABLE predictions (  
    id SERIAL PRIMARY KEY,  
    fire_id INTEGER REFERENCES fire_detections(id),  
    model_version TEXT,  
    prediction TEXT,  
    confidence_score INTEGER,  
    created_at TIMESTAMP DEFAULT NOW()  
);
```

3.5 Deployment (Docker)

File: **docker-compose.yml**

```
yaml
```

```
version: '3.8'

services:
  # Database
  postgres:
    image: postgis/postgis:15-3.3
    environment:
      POSTGRES_DB: wildfire_db
      POSTGRES_USER: wildfire_user
      POSTGRES_PASSWORD: secure_password
    volumes:
      - postgres_data:/var/lib/postgresql/data
    ports:
      - "5432:5432"

  # Backend API
  api:
    build: ./api
    command: uvicorn main:app --host 0.0.0.0 --port 8000
    volumes:
      - ./api:/app
      - ./outputs:/app/outputs
    ports:
      - "8000:8000"
    depends_on:
      - postgres
    environment:
      DATABASE_URL: postgresql://wildfire_user:secure_password@postgres:5432/wildfire_db

  # Frontend
  frontend:
    build: ./frontend
    command: npm run dev
    volumes:
      - ./frontend:/app
      - /app/node_modules
    ports:
      - "3000:3000"
    depends_on:
      - api

  # Nginx reverse proxy
  nginx:
```

```
image: nginx:alpine
volumes:
  - ./nginx.conf:/etc/nginx/nginx.conf
ports:
  - "80:80"
depends_on:
  - frontend
  - api

volumes:
  postgres_data:
```

Deploy:

```
bash
docker-compose up -d
```

3.6 Features to Implement

Core Features:

- Real-time fire detection dashboard
- Historical fire data visualization
- Batch prediction upload (CSV)
- Interactive map with filters
- Model performance metrics
- Export predictions as GeoJSON

Advanced Features:

- User authentication (Firebase/Auth0)
- Custom model training interface
- Alert system for high-confidence fires
- Time-series analysis and trends
- Mobile app (React Native)
- API rate limiting and authentication

3.7 Project Structure

```
wildfire-website/
  └── api/
    └── main.py      # FastAPI application
```

```
|- models/          # ML model files  
|- database.py     # Database connection  
|- schemas.py      # Pydantic models  
|- requirements.txt  
  
|- frontend/  
|   |- pages/  
|       |- index.js    # Home page  
|       |- dashboard.js # Analytics dashboard  
|       |- upload.js    # Batch upload  
|   |- components/  
|       |- Map.js       # Interactive map  
|       |- PredictionForm.js  
|       |- StatsPanel.js  
|   |- styles/  
|       |- package.json  
  
|- database/  
|   |- init.sql      # Database initialization  
  
|- docker-compose.yml  
|- nginx.conf  
|- README.md
```

Implementation Checklist

Week 1: Local Testing & Optimization

- Set up development environment
- Run full pipeline on your data
- Tune hyperparameters
- Document model performance
- Create sample predictions

Week 2: Backend Development

- Set up FastAPI project
- Implement prediction endpoints
- Add database integration
- Write API tests
- Deploy locally with Docker

Week 3: Frontend Development

- Set up Next.js project
- Create prediction form
- Integrate Leaflet map
- Connect to API
- Add error handling

Week 4: Integration & Deployment

- End-to-end testing
- Performance optimization
- Documentation
- Deploy to cloud (AWS/GCP/Heroku)
- Set up CI/CD pipeline

Week 5: Polish & Launch

- Add analytics
- Create demo video
- Write blog post
- Open-source release
- Community engagement

Success Metrics

Technical:

- API response time < 200ms
- Model accuracy > 85%
- 99.9% uptime
- Support 1000+ concurrent users

Community:

- 100+ GitHub stars in first month
- 10+ contributors
- 5+ forks
- Active Discord/Slack community

Resources for Website Development

Learning:

- [FastAPI Tutorial](#)
- [Next.js Documentation](#)
- [React Leaflet Examples](#)
- [Docker Compose Guide](#)

Hosting Options:

1. Free Tier Options:

- Frontend: Vercel, Netlify
- Backend: Railway, Render
- Database: Supabase, ElephantSQL

2. Production Options:

- AWS (EC2 + RDS + S3)
- Google Cloud Platform
- DigitalOcean

Getting Help

Stuck on Local Testing?

- Check `test_script.py` output for specific errors
- Review logs in `test_pipeline.log`
- Verify data files with `ls -lh *.geojson *.nc *.tif`

Stuck on Website Development?

- Join our Discord server (create one!)
- Open GitHub issues for bugs
- Check Stack Overflow for common problems

Next Steps

1. **Right now:** Run `python test_script.py` to validate your setup
2. **Today:** Execute full pipeline and review predictions

3. **This week:** Start FastAPI backend

4. **Next week:** Begin frontend development

5. **This month:** Deploy public beta!

You're ready to build something amazing! 