

## THE PROBLEM

# 55 Million Potholes. \$26.5B in Damage. Most Never Reported.

Reporting a civic issue takes 15–20 minutes of forms, phone calls, and wrong departments. Most citizens just give up.

**Snap it. Report it. Fix it.**

**55M**

Potholes on US roads

**\$26.5B**

Annual vehicle damage

**\$117B**

Annual infrastructure funding  
gap

**15–20**

Minutes to file a complaint  
today

### Issues covered:

Potholes

Cracked Pavement

Road Debris

Broken Signs

Faded Markings

# One Photo. Ten Seconds. Filed.

01

## Citizen Takes Photo

Open app, tap Report, snap or upload a photo of any road issue

02

## EfficientNetV2 Classifies

Custom model trained on 14,000+ civic images — 80% confidence threshold before proceeding

03

## Gemini Vision Generates

AI writes the complaint — cause, fix time, department routing. Only called if issue detected.

04

## Complaint Filed + Pinned

Saved to database, routed to the right city department, pinned live on the incident map

### Duplicate Detection

Flags if same issue reported within 100m

### Live Incident Map

Color-coded pins by severity across the city

### Department Routing

Auto-routes to Roads, Sanitation, Electrical

### Admin Dashboard

Real-time analytics, priority queue, status tracking

## TRAINING DATASET

Six Classes. 7,200 Images. Balanced Set.

7,200

Total Images

6

Classes

80/10/10

Train/Val/Test

Potholes

~1,200 images



Cracked Pavement

~1,200 images



Road Debris / Obstruction

~1,200 images



Broken Road Signs

~1,200 images



Faded Lane Markings

~1,200 images



Normal Road ✓

~1,200 images



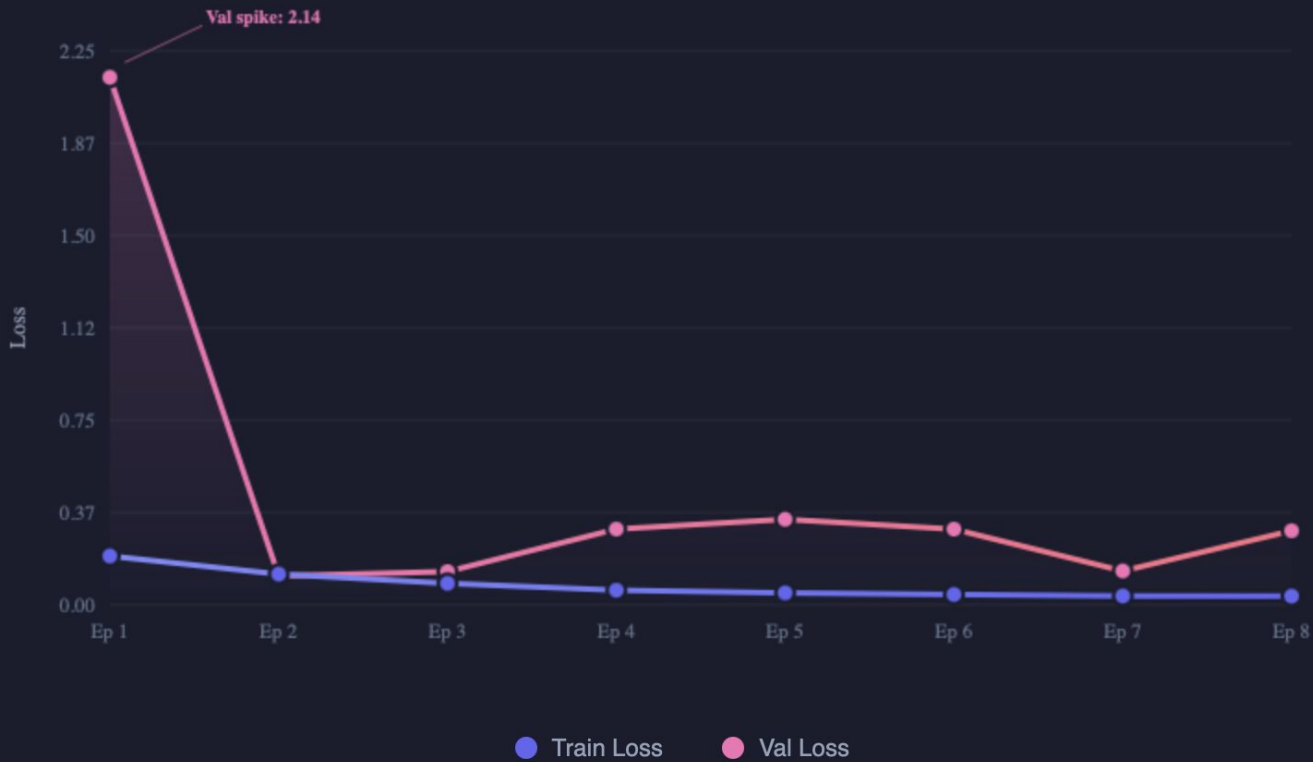
Source: [Roboflow Universe](#):  
Multiple road defect datasets merged, **deduplicated & cleaned**

Synthetic data:  
<100 bing search & manually annotated images for each class

Balanced to ~1,200 imgs/class  
Corrupt & undersized files removed before training

# Training & Validation Loss

Loss progression over 8 epochs



TEST ACCURACY



~ 800 images



## Sample Predictions — Green border = Correct

✓ True: broken\_road\_signs  
Pred: broken\_road\_Signs (100.00%)



✓ True: cracked\_pavement  
Pred: cracked\_pavement (73.42%)



✓ True: faded\_lane\_markings  
Pred: faded\_lane\_markings (100.00%)



✓ True: normal\_road  
Pred: normal\_road (100.00%)



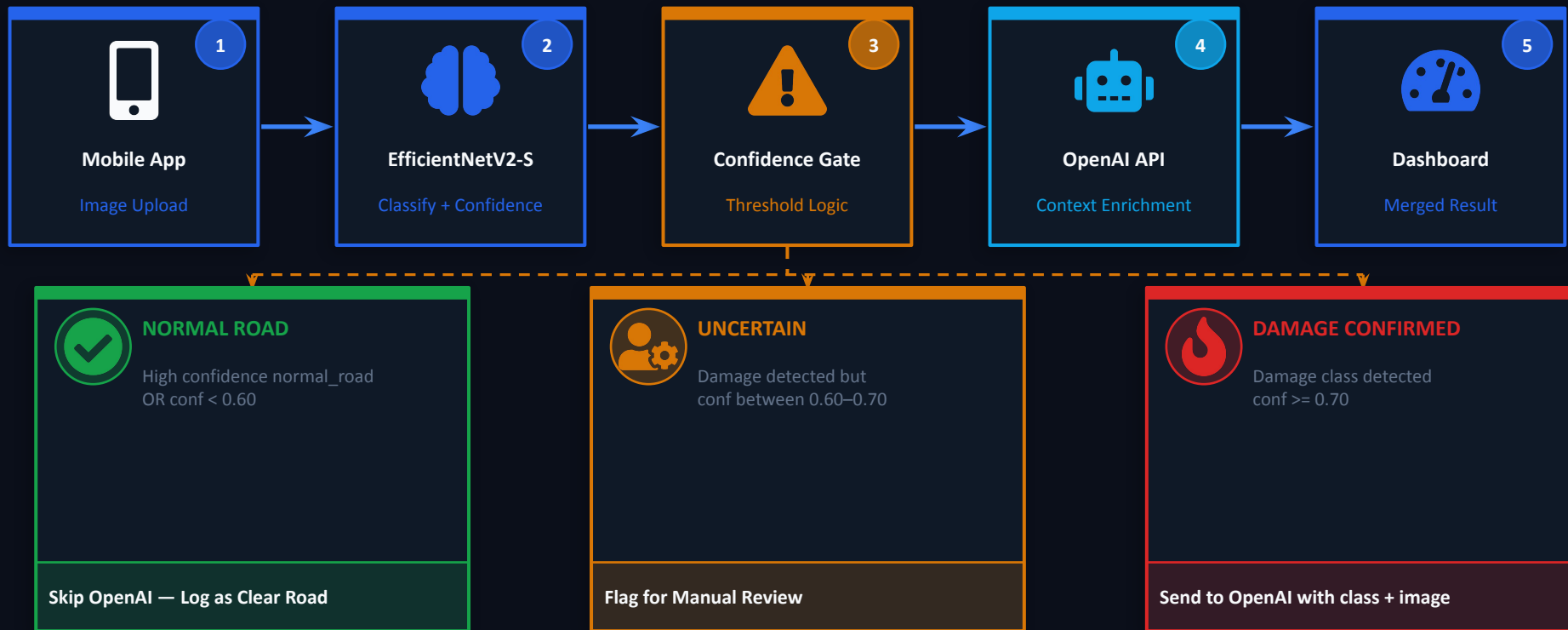
✓ True: potholes  
Pred: potholes (99.99%)



✓ True: road\_debris\_obstruction  
Pred: road\_debris\_obstruction (100.00%)



# Inference Pipeline — End-to-End Flow



# OpenAI Enrichment — Inputs & Outputs

## INPUT TO OPENAI



### Original Road Image

Full-resolution photo from mobile app upload



### Classifier Result

Damage class (e.g. pothole, crack, subsidence)



### Confidence Score

Numeric score  $\geq 0.70$  from EfficientNetV2-S

*The classifier's damage type is passed directly in the prompt for grounded, context-aware analysis.*



enriches

## OPENAI RETURNS



### Title

Severe pothole cluster on carriageway



### Description

Multiple deep potholes posing structural risk



### Severity

High — immediate action required



### Likely Cause

Water ingress + freeze-thaw cycle



### Est. Fix Time

4–6 hours, full patching crew



### Department

Roads & Infrastructure Division



# Confidence Routing — Decision Thresholds



## Zone 1 — conf < 0.60

Classified as normal road

OpenAI call is SKIPPED

Logged as: Clear Road

No issue ticket created



## Zone 2 — 0.60 to 0.70

Damage possible but uncertain

Flagged for manual review

Human inspector assigned

Issue held pending verification



## Zone 3 — conf >= 0.70

Damage class confirmed

Image + class sent to OpenAI

Full enrichment triggered

Auto-filed to dashboard