QUANTUM WOLF

DATA INTELLIGENCE RESEARCH LAB

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NAME: SURIYA E

Predictive Corrosion Mapping in Underground Pipelines

1. Problem Statement

Underground pipelines, which transport water and gas, are critical infrastructure components. However, hidden corrosion beneath insulation is a major issue, causing 12% of leaks and resulting in \$50 billion in repair costs annually. Current manual inspection methods fail to detect 70% of early-stage corrosion, leading to unexpected failures, environmental hazards, and significant financial losses. With 70% of U.S. pipelines over 50 years old, there is an urgent need for a predictive system to identify corrosion hotspots early and prevent leaks.

2. Solution

The project aims to predict corrosion hotspots using **ground-penetrating radar (GPR)**, **soil sensors**, and **machine learning models**. The solution involves:

- Collecting and organizing data from GPR, soil pH, and leak history.
- Cleaning and preprocessing the data.
- Engineering features like corrosion rate and humidity index.
- Predicting corrosion using multiple linear regression.
- Performing risk analysis and mapping corrosion severity using heatmaps.
- Scheduling preventive maintenance using Monte Carlo simulations.
- Generating a visual dashboard for performance reporting.

3. Working Process of Underground Pipeline Corrosion Prediction

Step 1: Data Collection

• Configuration:

- o Open the Excel file Corrosion_Prediction.xlsx.
- Create separate sheets for:
 - GPR Data: Columns for Location, GPR Signal Strength, Moisture Levels.
 - Soil pH Data: Columns for Location, Soil pH, Temperature.
 - Leak History: Columns for Location, Leak Incidents, Date.
 - Weights: Columns for Variables (Soil pH, Moisture, Temperature, GPR
 Signal) and their assigned weights.

Purpose:

o Organize raw data into structured formats for easy access and analysis.

Expected Output:

 A well-organized dataset with separate sheets for each data type, ready for cleaning and analysis.

Step 2: Data Cleaning

• Configuration:

- Select the dataset (e.g., Soil pH Data).
- o Go to Data → Remove Duplicates.
- For missing values, use:

=IFERROR(A2, AVERAGE(A:A))

o For outliers, use Z-Score or IQR methods (optional).

Purpose:

 Ensure data accuracy by removing duplicates, handling missing values, and managing outliers.

• Expected Output:

 Clean datasets free from duplicates, missing values, and outliers, ensuring reliable analysis.

| .1 | | | | |
|----------------|---------|----------|-------------|------------|
| _ A | В | С | D | Е |
| 1 Location | Soil_pH | Moisture | Temperature | GPR_Signal |
| 2 Location_1 | 6.62 | 17.41 | 21.54 | 0.71 |
| 3 Location_2 | 8.35 | 31.68 | 21.17 | 0.82 |
| 4 Location_3 | 7.70 | 44.92 | 37.66 | 0.33 |
| 5 Location_4 | 7.30 | 39.29 | 21.24 | 0.66 |
| 6 Location_5 | 5.97 | 42.26 | 21.80 | 0.61 |
| 7 Location_6 | 5.97 | 36.35 | 33.98 | 0.85 |
| 8 Location_7 | 5.67 | 37.69 | 26.24 | 0.92 |
| 9 Location_8 | 8.10 | 43.97 | 34.42 | 0.11 |
| 10 Location_9 | 7.30 | 19.99 | 16.63 | 0.71 |
| 11 Location_10 | 7.62 | 29.58 | 27.19 | 0.15 |
| 12 Location_11 | 5.56 | 18.85 | 15.84 | 0.59 |
| 13 Location_12 | 8.41 | 49.51 | 16.57 | 0.36 |
| 14 Location_13 | 8.00 | 47.76 | 37.66 | 0.38 |
| 15 Location_14 | 6.14 | 11.58 | 18.48 | 0.42 |
| 16 Location_15 | 6.05 | 38.22 | 28.31 | 0.66 |
| 17 Location_16 | 6.05 | 47.01 | 25.28 | 0.40 |
| 18 Location_17 | 6.41 | 17.22 | 23.68 | 0.76 |
| 19 Location_18 | 7.07 | 32.72 | 37.50 | 0.46 |
| 20 Location_19 | 6.80 | 46.62 | 15.55 | 0.16 |
| 21 Location_20 | 6.37 | 11.36 | 31.59 | 0.81 |
| 22 Location_21 | 7.34 | 37.90 | 39.08 | 0.36 |
| 23 Location_22 | 5.92 | 21.89 | 29.00 | 0.49 |
| 24 Location_23 | 6.38 | 46.98 | 38.42 | 0.72 |
| 25 Location 24 | 6.60 | 48.84 | 16.31 | 0.40 |

Step 3: Feature Engineering

Add New Columns for Feature Engineering

• Configuration:

o Add 3 new columns in the GPR Data Sheet:

• Column F: Corrosion Risk Factor

• Column G: Normalized GPR Signal

• Column H: Severity Score

Purpose:

 Create new features to improve the predictive model's accuracy and provide deeper insights into corrosion risk.

1. Corrosion Risk Factor (F2)

• Configuration:

- o In Cell F2, input the formula:
 - =IF(AND(B3<6.5, C3>30, D3>35, E3<0.3), "High Risk", IF(AND(B3<7, C3>25, D3>30, E3<0.5), "Moderate Risk", "Low Risk"))
- Drag the fill handle down to apply the formula to all rows.

Purpose:

Classify locations as High Risk, Moderate Risk and Low Risk based on soil pH
 (<6.5), moisture (>30), temperature (>35), and GPR signal (<0.3).

2. Normalized GPR Signal (G2)

• Configuration:

o In Cell G2, enter the formula:

Drag the fill handle down to apply the formula to all rows.

Purpose:

 Normalize the GPR signal values to a range between 0 and 1 for better comparison and analysis.

3. Severity Score (H2)

• Configuration:

o In Cell H2, enter the formula:

=C2*D2*E2

Drag the fill handle down to apply the formula to all rows.

• Purpose:

 Calculate a **Severity Score** by multiplying moisture, temperature, and GPR signal values to quantify corrosion severity.

Expected Outputs

1. Corrosion Risk Factor:

 A column classifying locations as High Risk or Low Risk based on environmental and GPR signal conditions.

2. Normalized GPR Signal:

 A column with GPR signal values normalized to a 0-1 range for easier comparison.

3. Severity Score:

 A column with calculated Severity Scores, quantifying corrosion risk for each location.

| 4 | | | | | | | | |
|-----|------------|---------|----------|-------------|------------|-----------------------|-----------------------|--------------------------|
| | Location | Soil_pH | Moisture | Temperature | GPR_Signal | Corrosion Risk Factor | Normalized GPR Signal | Corrosion Severity Score |
| L L | ocation_1 | 6.62 | 17.41 | 21.54 | 0.71 | Low Risk | 0.672786768 | 264.51 |
| L | ocation_2 | 8.35 | 31.68 | 21.17 | 0.82 | Moderate Risk | 0.796901161 | 547.99 |
| 1 L | ocation_3 | 7.70 | 44.92 | 37.66 | 0.33 | Low Risk | 0.250088527 | 550.43 |
| 5 L | ocation_4 | 7.30 | 39.29 | 21.24 | 0.66 | Low Risk | 0.62490541 | 552.73 |
| L | ocation_5 | 5.97 | 42.26 | 21.80 | 0.61 | Low Risk | 0.571719018 | 566.19 |
| L L | ocation_6 | 5.97 | 36.35 | 33.98 | 0.85 | Low Risk | 0.833089792 | 1049.53 |
| L | ocation_7 | 5.67 | 37.69 | 26.24 | 0.92 | Moderate Risk | 0.90642683 | 905.54 |
| L | ocation_8 | 8.10 | 43.97 | 34.42 | 0.11 | Low Risk | 0.011515999 | 167.88 |
| 0 L | ocation_9 | 7.30 | 19.99 | 16.63 | 0.71 | Low Risk | 0.674105137 | 234.92 |
| 1 L | ocation_10 | 7.62 | 29.58 | 27.19 | 0.15 | Low Risk | 0.05123855 | 117.93 |
| 2 L | ocation_11 | 5.56 | 18.85 | 15.84 | 0.59 | Low Risk | 0.548806595 | 177.34 |
| 3 L | ocation_12 | 8.41 | 49.51 | 16.57 | 0.36 | Moderate Risk | 0.287294123 | 294.32 |
| 4 L | ocation_13 | 8.00 | 47.76 | 37.66 | 0.38 | Low Risk | 0.306458992 | 676.52 |
| 5 L | ocation_14 | 6.14 | 11.58 | 18.48 | 0.42 | Low Risk | 0.352691552 | 89.36 |
| 6 L | ocation_15 | 6.05 | 38.22 | 28.31 | 0.66 | Low Risk | 0.621319831 | 713.29 |
| 7 L | ocation_16 | 6.05 | 47.01 | 25.28 | 0.40 | Low Risk | 0.333762274 | 476.08 |
| 8 L | ocation_17 | 6.41 | 17.22 | 23.68 | 0.76 | Moderate Risk | 0.732848633 | 309.77 |
| 9 L | ocation_18 | 7.07 | 32.72 | 37.50 | 0.46 | Low Risk | 0.404317 | 569.32 |
| 0 L | ocation_19 | 6.80 | 46.62 | 15.55 | 0.16 | Low Risk | 0.067774069 | 117.06 |

Step 4: Corrosion Prediction Using Multiple Linear Regression

• Configuration:

o In the **Weights Sheet**, assign weights to variables:

• Soil pH: 0.3

Moisture: 0.4

Temperature: 0.2

• GPR Signal: 0.1

o In the **Corrosion Prediction Sheet**, calculate corrosion scores using:

=SUMPRODUCT(B2:E2, Weights!\$B\$2:\$B\$5)

Purpose:

o Predict corrosion likelihood using a weighted linear regression model.

• Expected Output:

 A column of corrosion scores for each location, indicating the likelihood of corrosion.

Step 5: Risk Analysis (Anomaly Detection)

• Configuration:

- Select the Corrosion Score column (F2:F100).
- o Go to Conditional Formatting → Highlight Cells Rules → Greater Than...
 - Set > 0.8 (High Risk in Red).
 - Set 0.5 0.8 (Moderate Risk in Yellow).
 - Set < 0.5 (Low Risk in Green).

• Purpose:

o Identify and highlight high-risk corrosion areas for immediate attention.

• Expected Output:

 A visually formatted column showing risk levels (High, Moderate, Low) for each location.

| Location | Soil_pH | Moisture | Temperature | GPR_Signal | Corrosion Risk Factor | Normalized GPR Signal | Corrosion Severity Score |
|-------------|---------|----------|-------------|------------|-----------------------|-----------------------|--------------------------|
| Location_1 | 6.62 | 17.41 | 21.54 | 0.71 | Low Risk | 0.672786768 | 264.53 |
| Location_2 | 8.35 | 31.68 | 21.17 | 0.82 | Moderate Risk | 0.796901161 | 547.99 |
| Location_3 | 7.70 | 44.92 | 37.66 | 0.33 | Low Risk | 0.250088527 | 550.43 |
| Location_4 | 7.30 | 39.29 | 21.24 | 0.66 | Low Risk | 0.62490541 | 552.73 |
| Location_5 | 5.97 | 42.26 | 21.80 | 0.61 | Low Risk | 0.571719018 | 566.19 |
| Location_6 | 5.97 | 36.35 | 33.98 | 0.85 | Low Risk | 0.833089792 | 1049.53 |
| Location_7 | 5.67 | 37.69 | 26.24 | 0.92 | Moderate Risk | 0.90642683 | 905.54 |
| Location_8 | 8.10 | 43.97 | 34.42 | 0.11 | Low Risk | 0.011515999 | 167.88 |
| Location_9 | 7.30 | 19.99 | 16.63 | 0.71 | Low Risk | 0.674105137 | 234.92 |
| Location_10 | 7.62 | 29.58 | 27.19 | 0.15 | Low Risk | 0.05123855 | 117.93 |
| Location_11 | 5.56 | 18.85 | 15.84 | 0.59 | Low Risk | 0.548806595 | 177.34 |
| Location_12 | 8.41 | 49.51 | 16.57 | 0.36 | Moderate Risk | 0.287294123 | 294.32 |
| Location_13 | 8.00 | 47.76 | 37.66 | 0.38 | Low Risk | 0.306458992 | 676.52 |
| Location_14 | 6.14 | 11.58 | 18.48 | 0.42 | Low Risk | 0.352691552 | 89.36 |
| Location_15 | 6.05 | 38.22 | 28.31 | 0.66 | Low Risk | 0.621319831 | 713.29 |
| Location_16 | 6.05 | 47.01 | 25.28 | 0.40 | Low Risk | 0.333762274 | 476.08 |
| Location_17 | 6.41 | 17.22 | 23.68 | 0.76 | Moderate Risk | 0.732848633 | 309.77 |
| Location_18 | 7.07 | 32.72 | 37.50 | 0.46 | Low Risk | 0.404317 | 569.32 |
| Location_19 | 6.80 | 46.62 | 15.55 | 0.16 | Low Risk | 0.067774069 | 117.06 |

Step 6: Corrosion Severity Mapping (Heatmap)

• Configuration:

- o Select the entire Corrosion Score Column.
- o Go to Home → Conditional Formatting → Color Scale.
- o Choose Red-Yellow-Green Scale.

Purpose:

o Visualize corrosion severity across all locations using a color-coded heatmap.

• Expected Output:

 A heatmap showing corrosion severity, with high-risk areas in red and low-risk areas in green.

| | А | В | С | D | E | F | G | н | ı |
|--------|----------|---------|----------|-------------|------------|-----------------------|-----------------------|--------------------------|--------------------------|
| 1 | Location | Soil_pH | Moisture | Temperature | GPR_Signal | Corrosion Risk Factor | Normalized GPR Signal | Corrosion Severity Score | Corrosion Severity Score |
| 2 Loc | ation_1 | 6.62 | 17.41 | 21.54 | 0.71 | Low Risk | 0.672786768 | 264.51 | 264.51 |
| 3 Loc | ation_2 | 8.35 | 31.68 | 21.17 | 0.82 | Moderate Risk | 0.796901161 | 547.99 | 547.99 |
| 4 Loc | ation_3 | 7.70 | 44.92 | 37.66 | 0.33 | Low Risk | 0.250088527 | 550.43 | 550.43 |
| | ation_4 | 7.30 | 39.29 | 21.24 | 0.66 | Low Risk | 0.62490541 | 552.73 | 552.73 |
| 6 Loc | ation_5 | 5.97 | 42.26 | 21.80 | 0.61 | Low Risk | 0.571719018 | 566.19 | 566.19 |
| 7 Loc | ation_6 | 5.97 | 36.35 | 33.98 | 0.85 | Low Risk | 0.833089792 | 1049.53 | 1049.53 |
| 8 Loc | ation_7 | 5.67 | 37.69 | 26.24 | 0.92 | Moderate Risk | 0.90642683 | 905.54 | 905.54 |
| 9 Loc | ation_8 | 8.10 | 43.97 | 34.42 | 0.11 | Low Risk | 0.011515999 | 167.88 | 167.88 |
| 10 Loc | ation_9 | 7.30 | 19.99 | 16.63 | 0.71 | Low Risk | 0.674105137 | 234.92 | 234.92 |
| 11 Loc | ation_10 | 7.62 | 29.58 | 27.19 | 0.15 | Low Risk | 0.05123855 | 117.93 | 117.93 |
| 12 Loc | ation_11 | 5.56 | 18.85 | 15.84 | 0.59 | Low Risk | 0.548806595 | 177.34 | 177.34 |
| 13 Loc | ation_12 | 8.41 | 49.51 | 16.57 | 0.36 | Moderate Risk | 0.287294123 | 294.32 | 294.32 |
| 14 Loc | ation_13 | 8.00 | 47.76 | 37.66 | 0.38 | Low Risk | 0.306458992 | 676.52 | 676.52 |
| 15 Loc | ation_14 | 6.14 | 11.58 | 18.48 | 0.42 | Low Risk | 0.352691552 | 89.36 | 89.36 |
| 16 Loc | ation_15 | 6.05 | 38.22 | 28.31 | 0.66 | Low Risk | 0.621319831 | 713.29 | 713.29 |
| 17 Loc | ation_16 | 6.05 | 47.01 | 25.28 | 0.40 | Low Risk | 0.333762274 | 476.08 | 476.08 |
| 18 Loc | ation_17 | 6.41 | 17.22 | 23.68 | 0.76 | Moderate Risk | 0.732848633 | 309.77 | 309.77 |
| 19 Loc | ation_18 | 7.07 | 32.72 | 37.50 | 0.46 | Low Risk | 0.404317 | 569.32 | 569.32 |
| 20 Loc | ation_19 | 6.80 | 46.62 | 15.55 | 0.16 | Low Risk | 0.067774069 | 117.06 | 117.06 |

Step 7: Preventive Maintenance Scheduling Using Monte Carlo Simulation

• Configuration:

- o Create a new sheet named **Monte Carlo Simulation**.
- o In Column A, copy all Location Names (Location_1 to Location_20).
- o In Column B, copy the Corrosion Score from the previous calculation.
- o In Cell C2, enter:

- Drag the formula down for 1000 rows for each location.
- o In Column D, calculate the Failure Probability:

=COUNTIF(C2:C1001, ">=0.8") / 1000

o Visualize the Monte Carlo Simulation using a Bar Chart or Histogram.

Purpose:

 Simulate 1000 scenarios of corrosion failure to predict failure probabilities and schedule preventive maintenance.

• Expected Output:

 A column of failure probabilities for each location and a visual chart showing failure patterns.

| | | | | | 1 | ı |
|----|----------|-------------|------------|--------------------------|------------------------|---------------------|
| 4 | С | D | E | F | G | Н |
| 1 | Moisture | Temperature | GPR_Signal | Corrosion Severity Score | Simulation (1000 Runs) | Failure Probability |
| 2 | 17.41 | 21.54 | 0.71 | 264.51 | 223.5 | 0.997 |
| 3 | 31.68 | 21.17 | 0.82 | 547.99 | 199.6 | 0.996 |
| 4 | 44.92 | 37.66 | 0.33 | 550.43 | 373.1 | 0.995 |
| 5 | 39.29 | 21.24 | 0.66 | 552.73 | 390.0 | 0.994 |
| 6 | 42.26 | 21.80 | 0.61 | 566.19 | 541.7 | 0.993 |
| 7 | 36.35 | 33.98 | 0.85 | 1049.53 | 608.6 | 0.992 |
| 8 | 37.69 | 26.24 | 0.92 | 905.54 | 604.5 | 0.991 |
| 9 | 43.97 | 34.42 | 0.11 | 167.88 | 18.9 | 0.99 |
| 10 | 19.99 | 16.63 | 0.71 | 234.92 | 84.7 | 0.989 |
| 11 | 29.58 | 27.19 | 0.15 | 117.93 | 80.2 | 0.988 |
| 12 | 18.85 | 15.84 | 0.59 | 177.34 | 31.7 | 0.987 |
| 13 | 49.51 | 16.57 | 0.36 | 294.32 | 233.7 | 0.986 |
| 14 | 47.76 | 37.66 | 0.38 | 676.52 | 281.0 | 0.985 |
| 15 | 11.58 | 18.48 | 0.42 | 89.36 | 83.3 | 0.984 |

Step 8: Performance Report (Graph Dashboard)

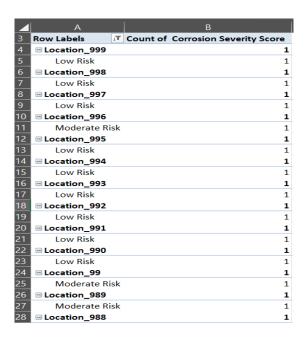
• Configuration:

- Pivot Table for Corrosion Severity:
 - Select your Corrosion Prediction Sheet.
 - Click on Insert → Pivot Table.
 - Choose "New Worksheet" and click OK.
 - Configure the Pivot Table:

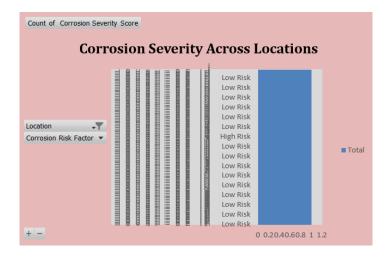
Rows: Location

Values: Corrosion Severity (set to "Count")

o **Output:** A Pivot Table summarizing corrosion severity.

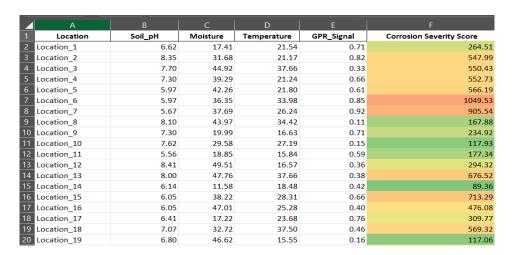


- o Bar Chart for High-Risk Locations:
 - Select the Pivot Table Data.
 - Click Insert → Bar Chart.
 - Format the chart:
 - High-Risk locations in Red.
 - Moderate-Risk locations in Orange.
 - Low-Risk locations in Green.
 - Add Title: "Corrosion Severity Across Locations".
- Output: A Bar Chart showing high-risk locations.



Heatmap for Corrosion Mapping:

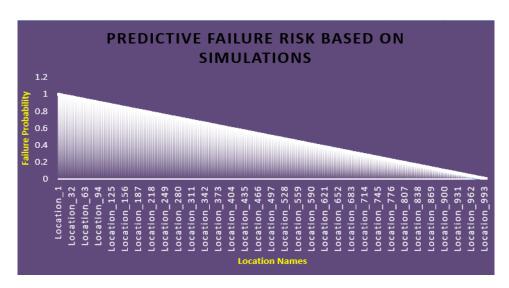
- Select the Corrosion Severity Column.
- Click Home → Conditional Formatting → Color Scales.
- Choose Red-Yellow-Green Scale.
- Output: A Heatmap for corrosion mapping.



Predictive Failure Chart (Monte Carlo Data):

- Go to Monte Carlo Simulation Sheet.
- Select Failure Probability Data.
- Click Insert → Line Chart.
- Customize:
 - X-axis = Location Names

- Y-axis = Failure Probability
- Add Title: "Predictive Failure Risk Based on Simulations".
- o **Output:** A Predictive Failure Chart based on Monte Carlo simulations.



4. Final Outcomes and Workflow

Final Outputs

Outcome 1: Identified high-risk corrosion hotspots using predictive modeling.

o A list of locations classified as High, Moderate, or Low Risk

Outcome 2: Generated a heatmap for visualizing **corrosion severity** across pipeline locations.

o A color-coded map showing corrosion severity across all locations.

Outcome 3: Developed a **preventive maintenance schedule** using Monte Carlo simulations.

o A schedule with failure probabilities for each location.

Outcome 4: Created a visual dashboard for real-time monitoring and reporting.

o A comprehensive dashboard for real-time monitoring and reporting.

Workflow:

Data Collection → 2. Data Cleaning → 3. Feature Engineering → 4. Corrosion
 Prediction → 5. Risk Analysis → 6. Severity Mapping → 7. Maintenance Scheduling →
 8. Performance Reporting.

5. Problems Faced During the Project

1. Heatmap Visualization:

- Difficulty in scaling the heatmap for large datasets.
- Addressed by using pivot tables and conditional formatting.

2. Monte Carlo Simulation:

- o High computational load for 1000 simulations.
- Optimized by using Excel's Data Table feature.

6. Conclusion

The **Predictive Corrosion Mapping in Underground Pipelines** project successfully addresses the critical issue of hidden corrosion in aging pipelines, which leads to leaks, environmental hazards, and significant financial losses. By leveraging **Excel-based tools**, **predictive modeling**, and **data visualization techniques**, the project provides a cost-effective and scalable solution for early detection and prevention of pipeline corrosion.

Impact:

- Cost Savings: Reduces annual repair costs by preventing leaks and addressing corrosion early.
- Improved Safety: Minimizes environmental and safety risks associated with pipeline failures.
- Extended Lifespan: Enhances the longevity of aging pipelines, 70% of which are over 50 years old in the U.S.
- Resource Optimization: Allocates maintenance resources efficiently, focusing on high-risk areas.