

CNN-based Deep Learning-Enabled Tool for Winter Roadway Condition Monitoring

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Introduction & Background

Introduction



RWIS Database:

Traffic Data

Surface Data

Camera Data

Atmospheric Data





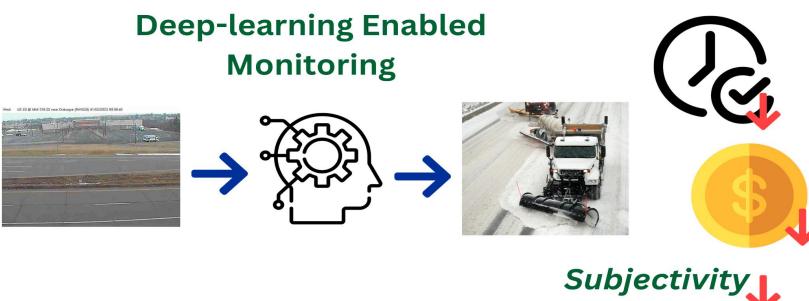
❖ Winter challenges road safety and mobility with snow and ice, necessitating effective maintenance strategies. While Road Weather Information Systems (RWIS) equipped with cameras have improved the monitoring of road conditions, the manual analysis of Road Surface Conditions (RSC) captured by these cameras is still time-inefficient and labour-intensive.

Speed 17%↓



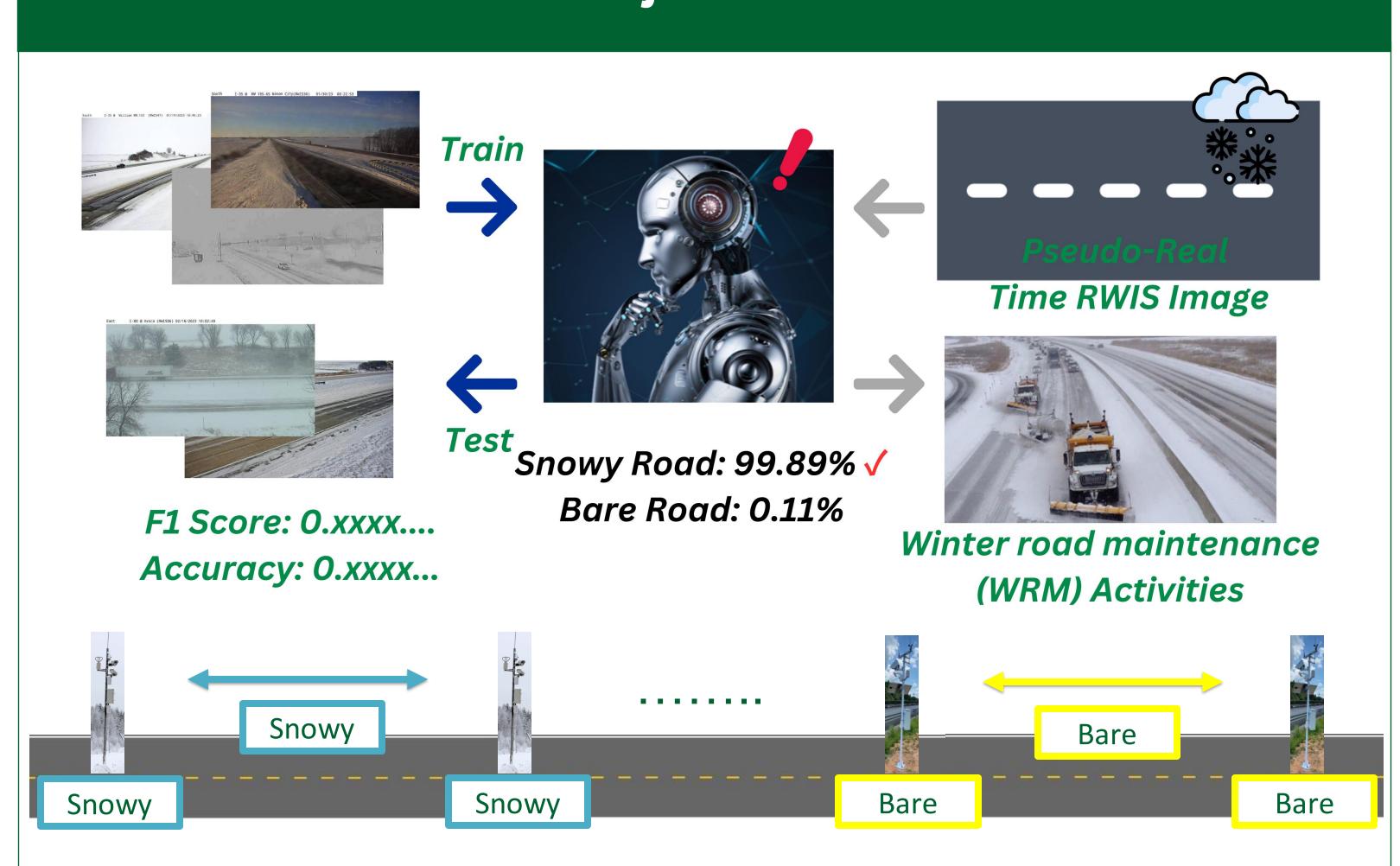
Background





Advances in artificial intelligence and computer vision allow for the automated monitoring of RSC through RWIS image analysis. This enables real-time monitoring and reduces labor costs, which are vital for improving road safety and efficiency during winter.

Objectives

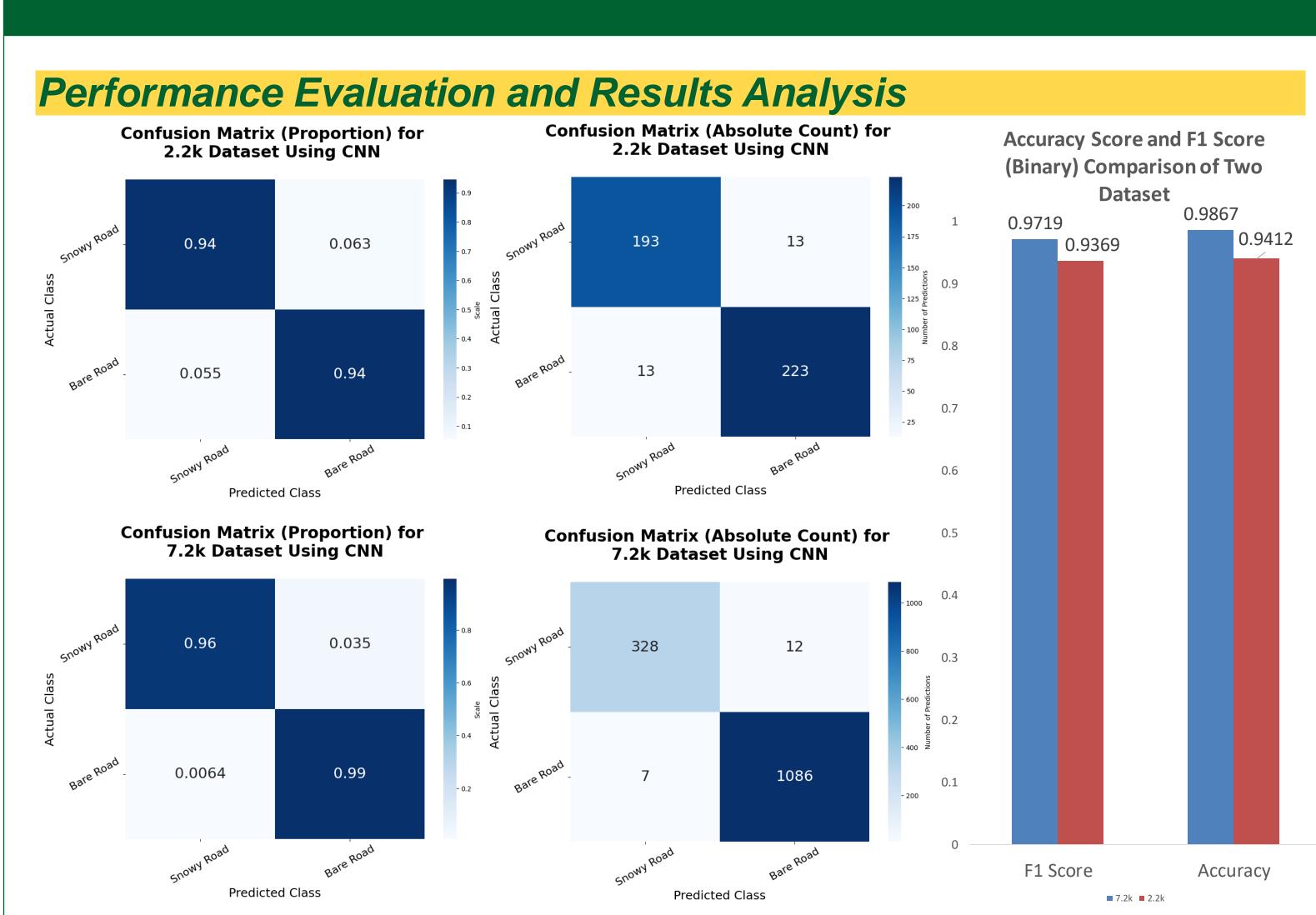


The objective is to improve winter road monitoring with Convolutional Neural Networks (CNN) to enable near real-time RSC monitoring by interpolating between RWIS points.

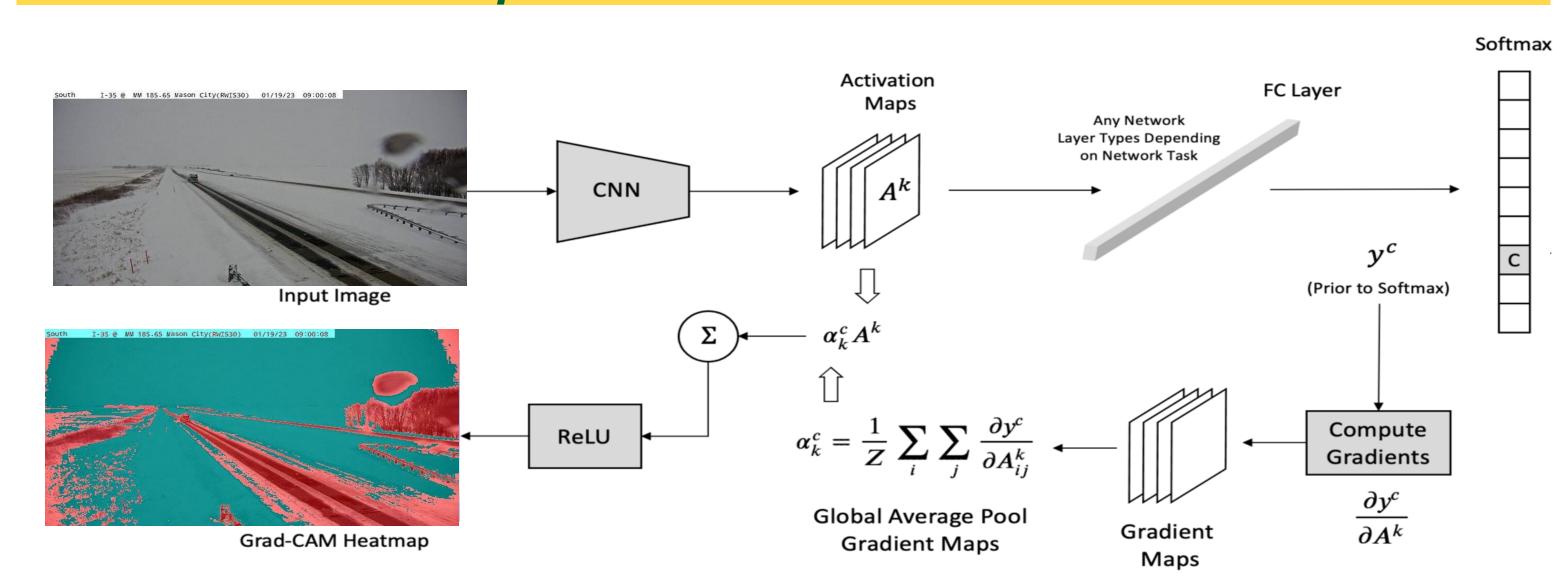
Methodology Study Area & Data Collection 10-year Average Snow (cm) in Iowa Iowa State university **Stationary RWIS Iowa RWIS** Dec 2022 – Feb 2023 Location Image Labelling Snowy Road Bare Road Comparison of Unbalanced and Balanced Image Datasets DISTRIBUTION OF ROAD SURFACE DISTRIBUTION OF ROAD SURFACE **CONDITIONS FOR 7.2K CONDITIONS FOR 2.2K** Snowy Road, 1717, 24% Bare Road **_1148, 52%** Snowy Road, 1061, 48% Bare Road, 5447, 76% Absolute Counts of Primary Class by Angle, Dataset 7.2k Absolute Counts of Primary Class by Angle, Dataset 2.2k Bare Road Bare Road Snowy Road Snowy Road Unbalanced 7,164 Images (7.2k) Balanced 2,200 Images (2.2k) 67 Individual Angles 17 Individual Angles Training & Testing with Convolutional Neural Networks (CNN) Load .npy Files in Split Data 80:20 for Generate and Export Load Image .npy Files using CPU Training/Testing **GPU Environment** Train in Flatten, Split Training Data Test Model Save Model and 80:20 for Training Performance using Dense, Dropout, and Validation Accuracy and F1 Dense Order Fully Connected Convolution Pooling Output Input

Feature Extraction

Results



Grad-CAM: Visual Explanations via Gradient-based Localization



Grad-CAM enhances CNN interpretability by highlighting image regions most influential in classification, such as snow or bare pavement, offering insights into model decisions and improvement areas.

Conclusions

❖ A CNN-based deep learning model was developed to automate RSC recognition from stationary RWIS images, demonstrating high accuracy in training and validation. Its success not only proves its precision in identifying RSC but also showcases its potential in advancing towards a pseudo-real-time road condition monitoring system, offering significant cost and labor reductions in road safety operations.

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Classification



