

# *Stressed about Distress? Calibrating Our Way to Quality in Automated Pavement Data Collection*

*51<sup>st</sup> Annual Mid-Atlantic Quality Assurance Workshop*



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AECOM

# Presentation Overview

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- Introduction
- High-Speed Collection and Automated Analysis - Review
- Data Quality and Challenges
- Case Study: DeIDOT Approach and Successes
- Future Trends and Recommendations

# Introduction

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- Kathy Keegan, P.E.
  - 23 years experience with high-speed data collection and distress analysis.
  - State, municipal, aviation



- Tim Miller, P.E.
  - 10 years experience with high-speed data collection and distress analysis.
  - State, municipal, aviation

# Data Collection Evolution



Manual Collection with Paper



Add Photos (LTPP Pasco)



Add GPS

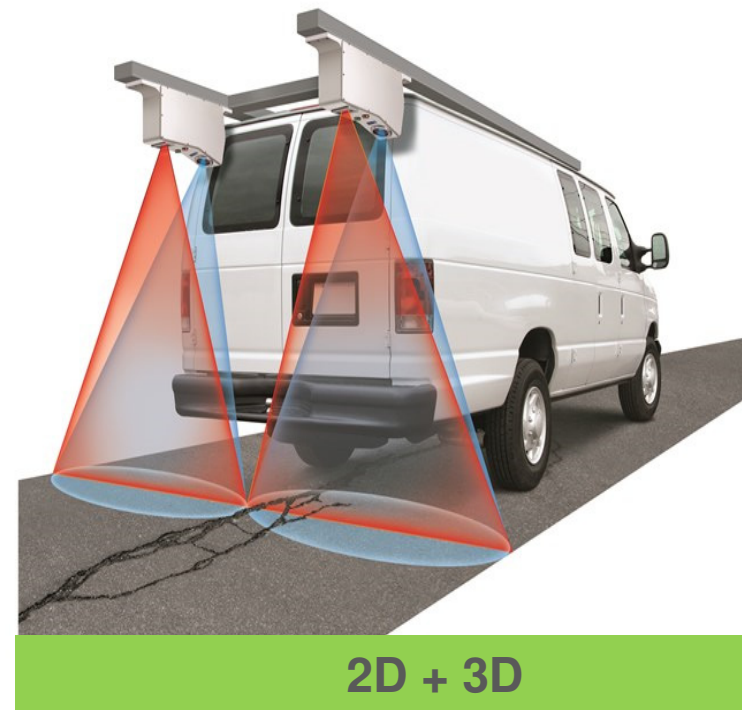
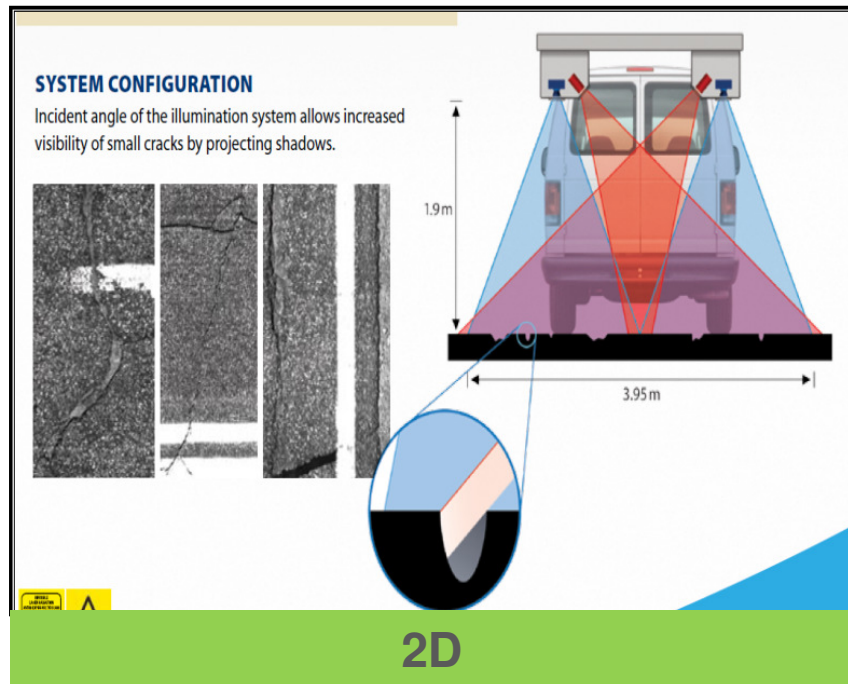


2D Photos (video-like)



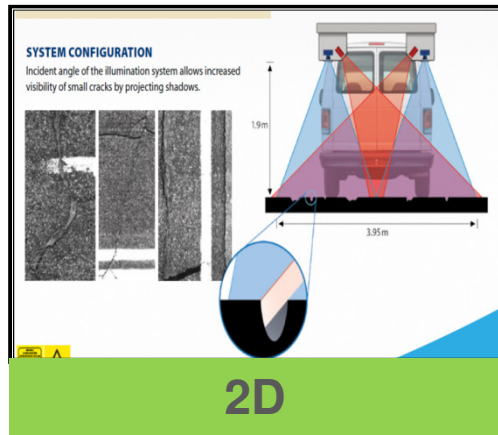
2D and 3D Imagery Combined

# High-Speed Data Collection

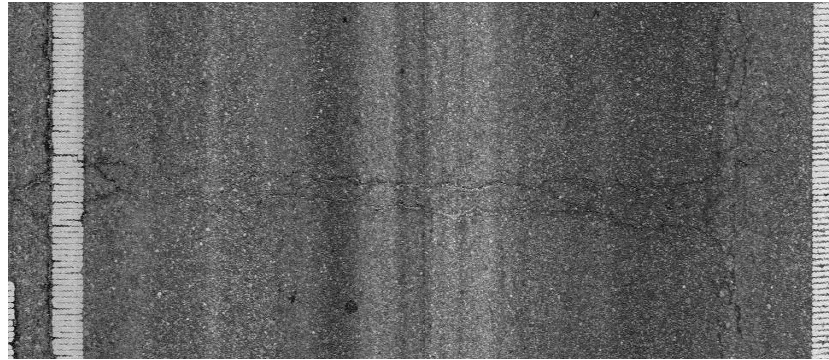




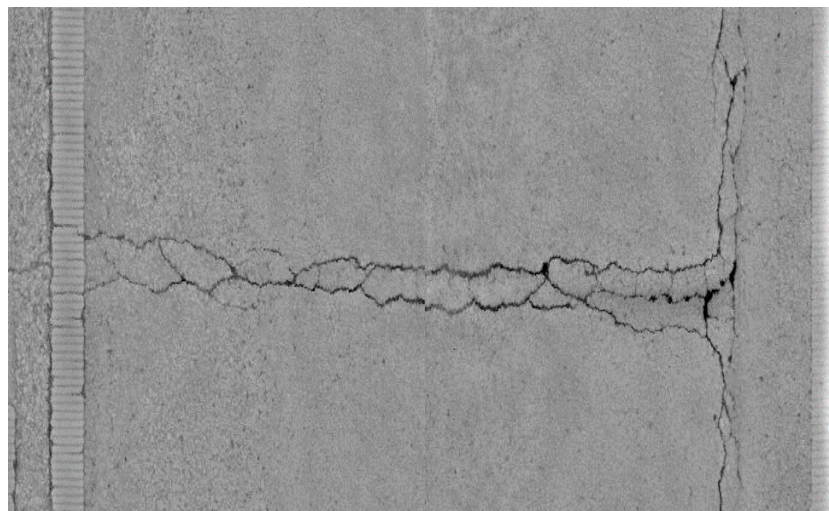
# High-Speed Data Collection



## Conventional 2D Imagery - LRIS



## High Definition 3D Imagery - LCMS







# High Speed Data Collection and Automated Analysis

# High-Speed Data Collection

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- HPMS
  - IRI, Rut, Fault, %Crack
- Collect additional data in support of broader planning
- State of Practice (FHWA 2015)

High-Speed	Self Collect	Vendor	Automated Distress
All	19	31	7

- ‘Semi-automated’



# Automated Analysis

- Analysis is driven by a ‘standard’
- Standards can vary!
- Manual
  - ‘Boots on ground’
  - Subjective
- Semi-Automated
  - Mix of Artificial Intelligence (AI) and subjective
- Automated
  - Fully AI
  - Less subjective
  - Accurate and Precise?





Route	Direction	From MP	To MP	PC Lon	Length	Degree	S	
80	W	128		-40.712	-111.801	134,951.52	0.083	0.042_f
80	W	127.91	127.862	-40.713	-111.802	2,147.91	0.055	2.672_f
80	W		127.79	-40.713	-111.803	216,357.82	0.073	-0.032_f
80	W	127.79	127.674	-40.714	-111.804	2,562.66	0.136	-2.242_f
80	W	127.674	127.556	-40.714	-111.805	5,191.77	0.031	-1.132_f
80	W		127.556	-40.714	-111.807	40,715	0.086	-2.622_f
80	W	127.556	127.438	-40.714	-111.809	40,715	0.245	0.2_f
80	W	127.438	127.320	-40.715	-111.813	40,714	0.124	-4.522_f
80	W	127.320	127.202	-40.714	-111.815	40,714	0.129	0.042_f

# Data Quality and Challenges

# Quality

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- What is Quality?
  - ‘Degree of Excellence’ = Accurate and Precise
- Quality Assurance?
  - ‘The maintenance of a desired level of quality in a service or product, especially by means of attention to every stage of the process of delivery or production.’





# Quality

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- Why it matters?
  - Extensive planning and budgeting done as a result of data collected.



# Quality Management

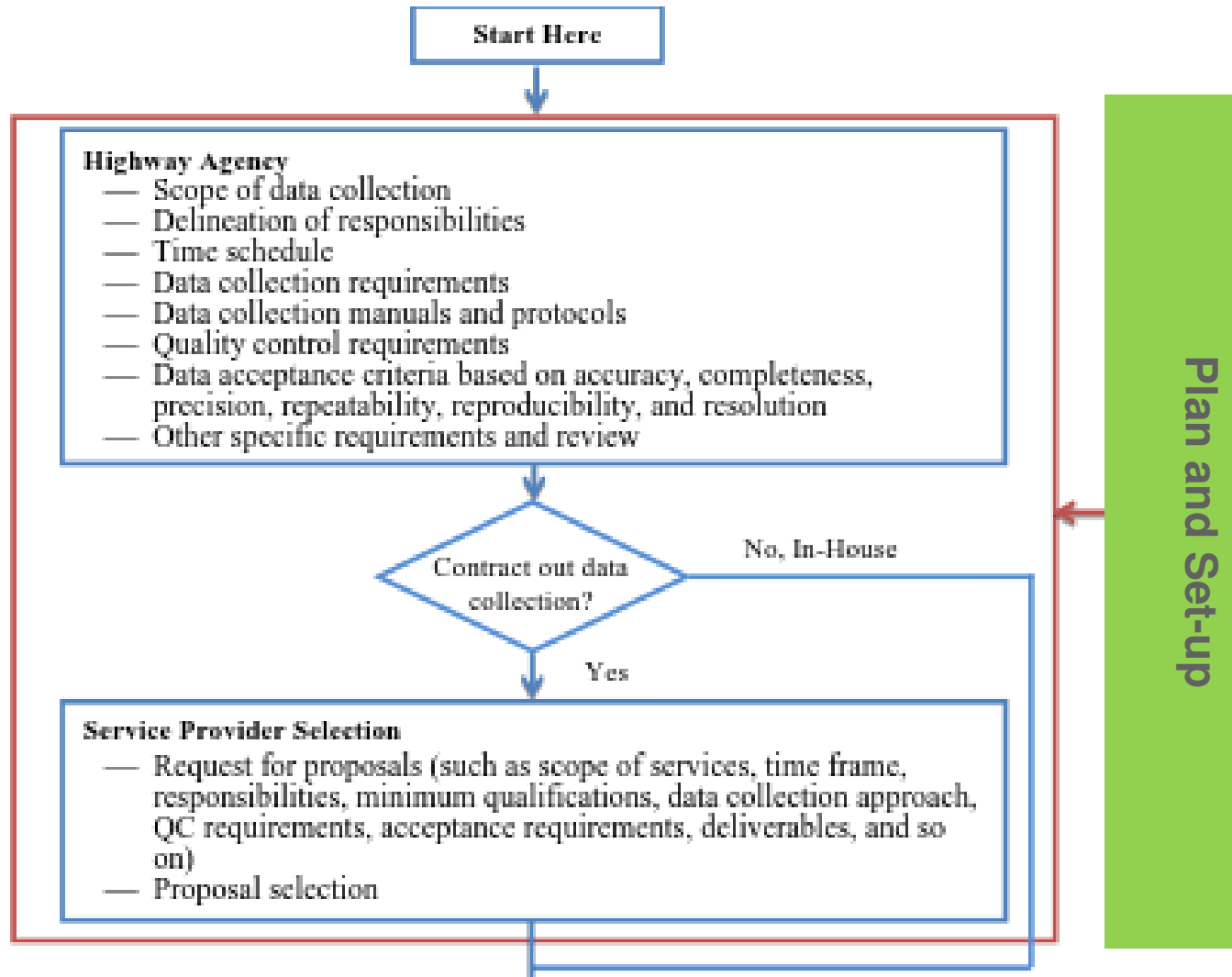
- Example:
  - FHWA
  - Individual State Protocols

## Practical Guide for Quality Management of Pavement Condition Data Collection



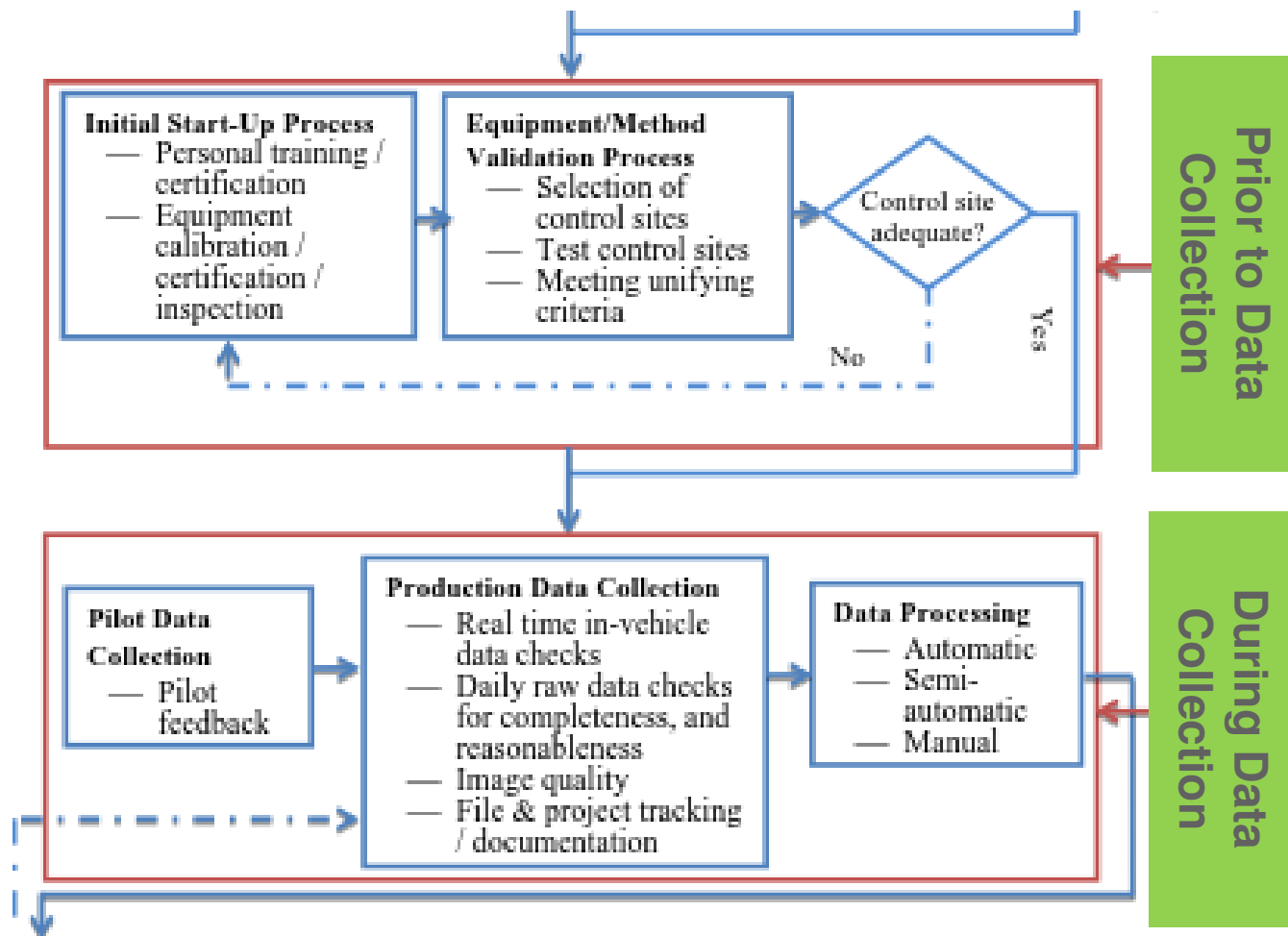
U.S. Department of Transportation  
Federal Highway Administration

# Quality Management

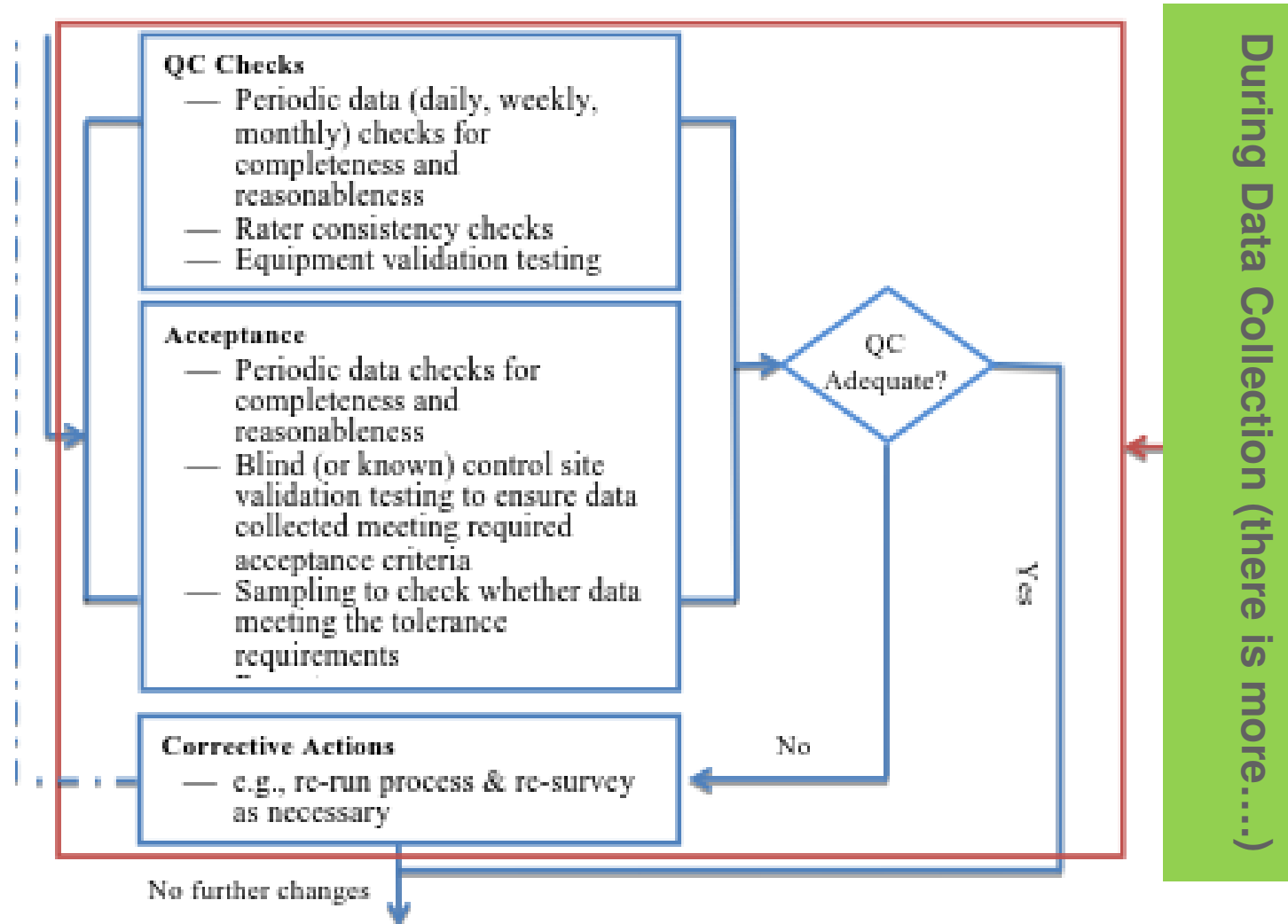




# Quality Management

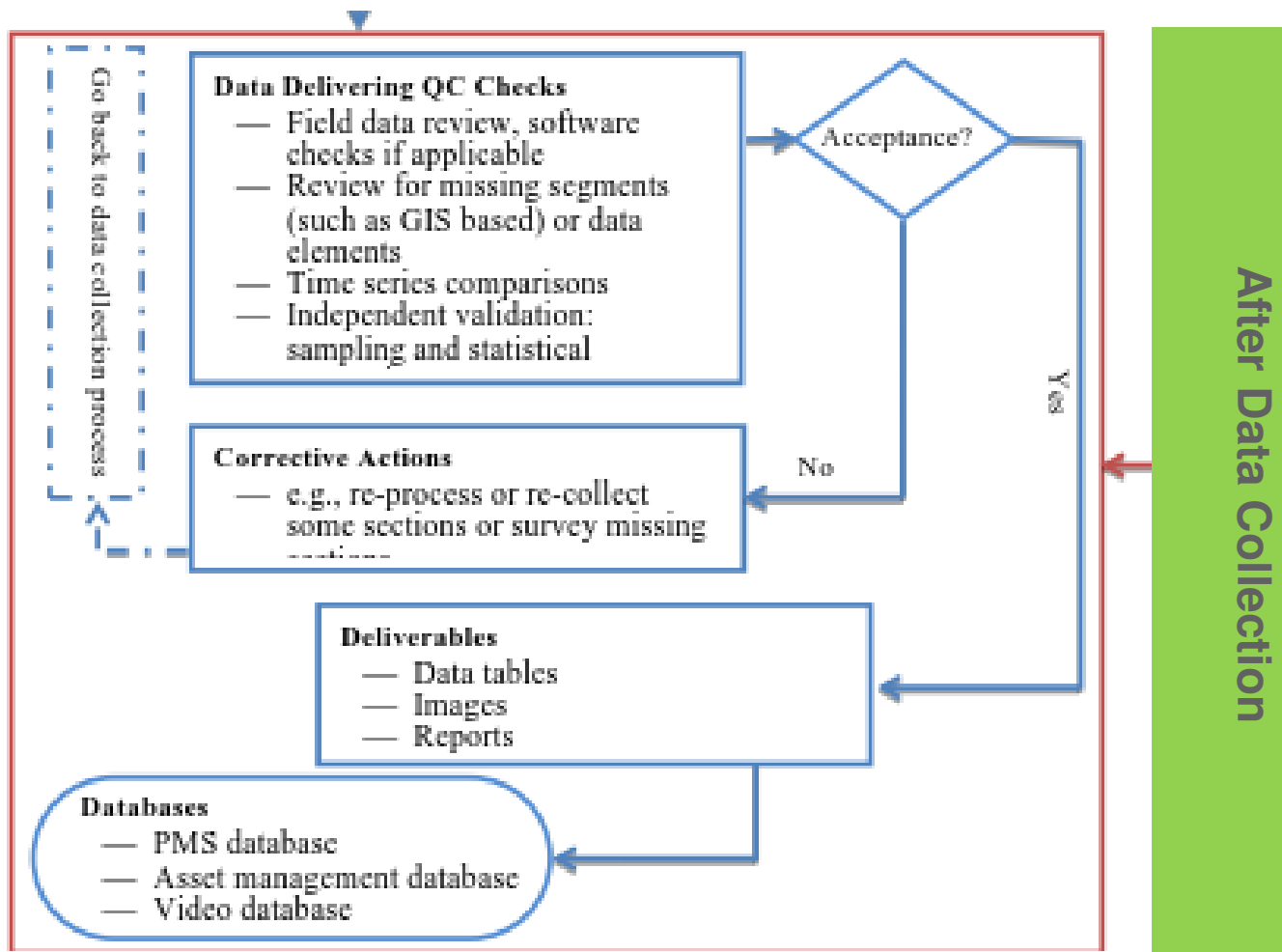


# Quality Management



During Data Collection (there is more...)

# Quality Management

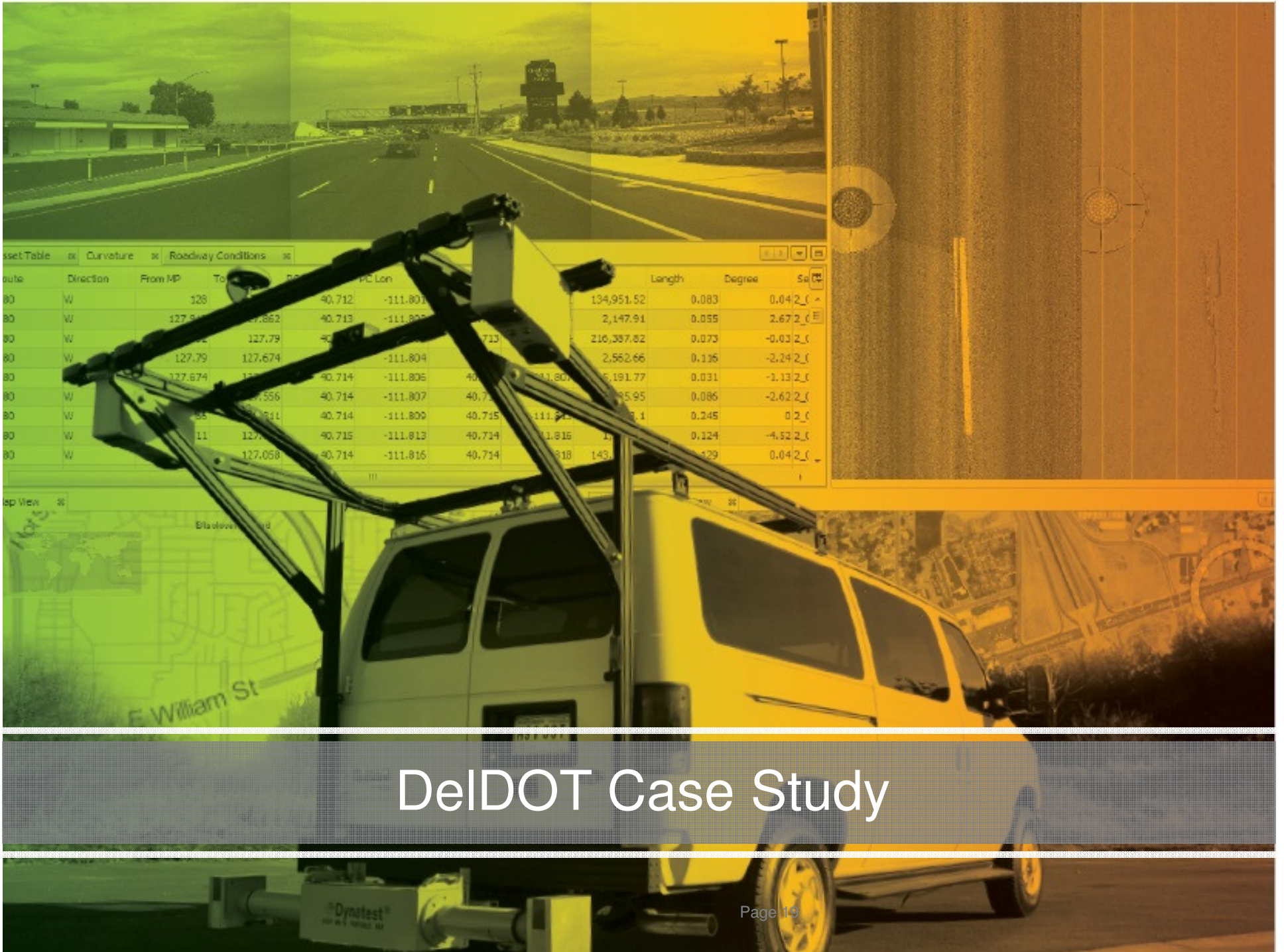




# Challenges

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- Variables
  - Equipment
  - Vendors
  - Collection Protocols
  - Analysis Protocols
  - Software / Analysis Tools
- Technology
  - Continues to advance, so should our approach.
- Interpretation of Data
  - Does resultant data match with expert opinion?
  - Does data yield the correct treatment / timing?



# DeIDOT Case Study

# Acknowledgements

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- DeIDOT: Sarah McDougall, Rhonda Lewis, Mike Beulah, Sanjay Kumar
- AECOM: Tim Miller, Kathy Keegan, Amir Arshadi, Tamim Khan, Mike Papakostas
- Kercher: Aaron Gerber, Qazi Aurangzeb, Eric Perrone
- Mandli: Mike Richardson, Celly Chrisinger, Larry Mattke, Ryan Well

# Project Motivation

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- Over decade: change in vendor, collection equipment, and survey method.
- Result: Overall Pavement Condition Index (OPC) values did not make sense (not following expected trends).
- Advancing technology means time for change in approach!

# Project Motivation

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- Key changes included:
- Transition from 3x3 matrix definitions to a data dictionary encompassing all distress types at all severity levels.
  - Now that we can easily quantify accurately, why don't we?

Severity/Extent	Low	Med	High
Low			✓
Med	✓		
High	✓		

- Undertake calibration exercises to align OPC values, treatment recommendations, and expectations.



# Project Objectives

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- Narrow the discrepancies between LCMS (3D)-based pavement distress data and network-level treatment selections.
- Select calibration sites and collect baseline measurements
- Reconcile differences in manual and automated surveys
- Adjust PMS index models

# Fatigue Cracking Matrix Definitions

- OPC influenced by dominant distress index severity only
- Issues in calculating the most representative OPC for a pavement section
- Limited number of distresses captured per pavement type
- Relies on estimated quantity

<b>Severity Level</b>	<b>Low Extent</b>	<b>Medium Extent</b>	<b>High Extent</b>
Low Severity	Ext: 1 - 9% (wheel path) Sev: Fine parallel hairline cracks	Ext: 10 - 25% Sev: Fine parallel hairline cracks	Ext: > 25% Sev: Fine parallel hairline cracks
Medium Severity	Ext: 1 - 9% (wheel path) Sev: Alligator crack pattern clearly developed	Ext: 10 - 25% Sev: Alligator crack pattern clearly developed	Ext: > 25% Sev: Alligator crack pattern clearly developed
High Severity	Ext: 1 - 9% (wheel path) Sev: Alligator crack pattern clearly developed with spalling and/or distortion	Ext: 10 - 25% Sev: Alligator crack pattern clearly developed with spalling and/or distortion	Ext: > 25% Sev: Alligator crack pattern clearly developed with spalling and/or distortion

# Data Dictionary Development

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- Defines all distress types and severity levels
- Defines methods of measurement for automated road rating data collection
- Includes four pavement types
- Utilizes actual extent measurements rather than discrete extent ranges (% estimates)

# Data Dictionary Revisions

Distress	Pavement Type				Measurement Type
	Asphalt	Composite	Surface Treated	PCC	
Fatigue Cracking	H	H	H		Square Feet
Transverse Cracking	H		R		Count
Block Cracking	H	H	R		Square Feet
Joint Reflective Cracking		H			Count
Edge Cracking			H		Linear Feet
Non-Wheel Path Longitudinal Cracking	R	R			Linear Feet
Patches/Potholes	H	R	R		Square Feet
Bleeding			H		Square Feet
Raveling	H	H	H		Square Feet
Crown/Cross-Slope			H		Percent
Slab Cracking				H	Slab Count
Joint Deterioration				H	Joint Count
Joint Seal Loss				H	Joint Count
Patch Deterioration				H	Square Feet
Alkali-Silica Reactivity (ASR)				H	Slab Count

Items denoted as 'H' have been used historically by DeIDOT, while items denoted with 'R' are revisions to the process.

# Data Progression

Distress Raw Data

% Extent of Distress

IDI Calculations

Structural Index

Non-structural Index

OPC  
Calculation

IRI, Rut

ASR



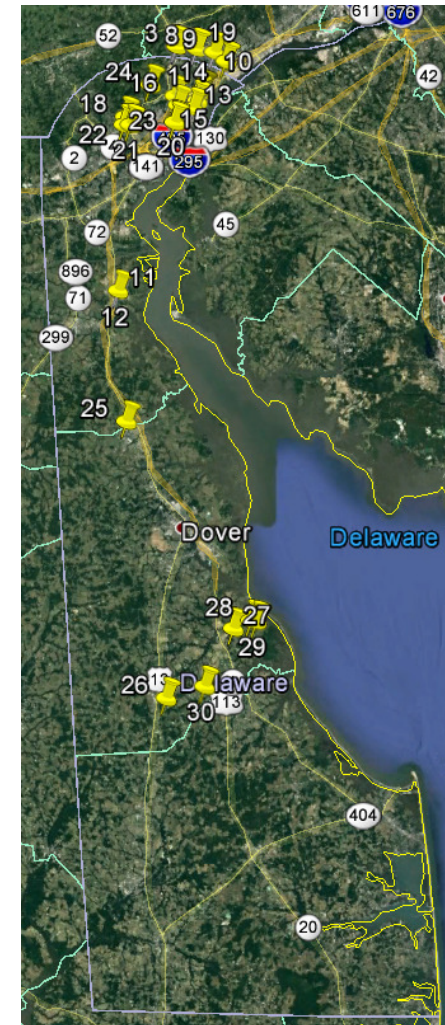
**Million  
Dollar  
Question!**

*What is the best type of treatment?*

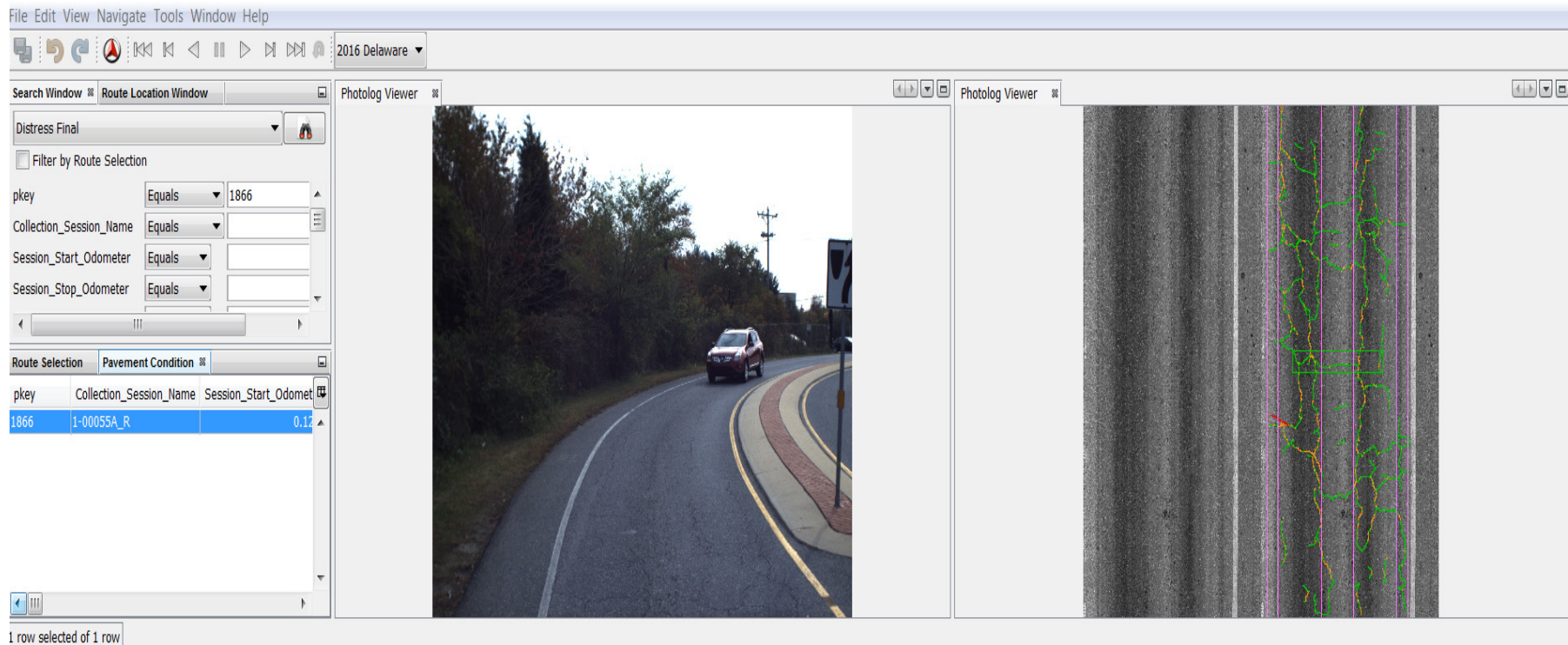


# Site Locations

- 30 sites selected for initial calibration
- 4 pavement types
- Range of pavement conditions
- Lengths range from 0.05 to 0.1 mi

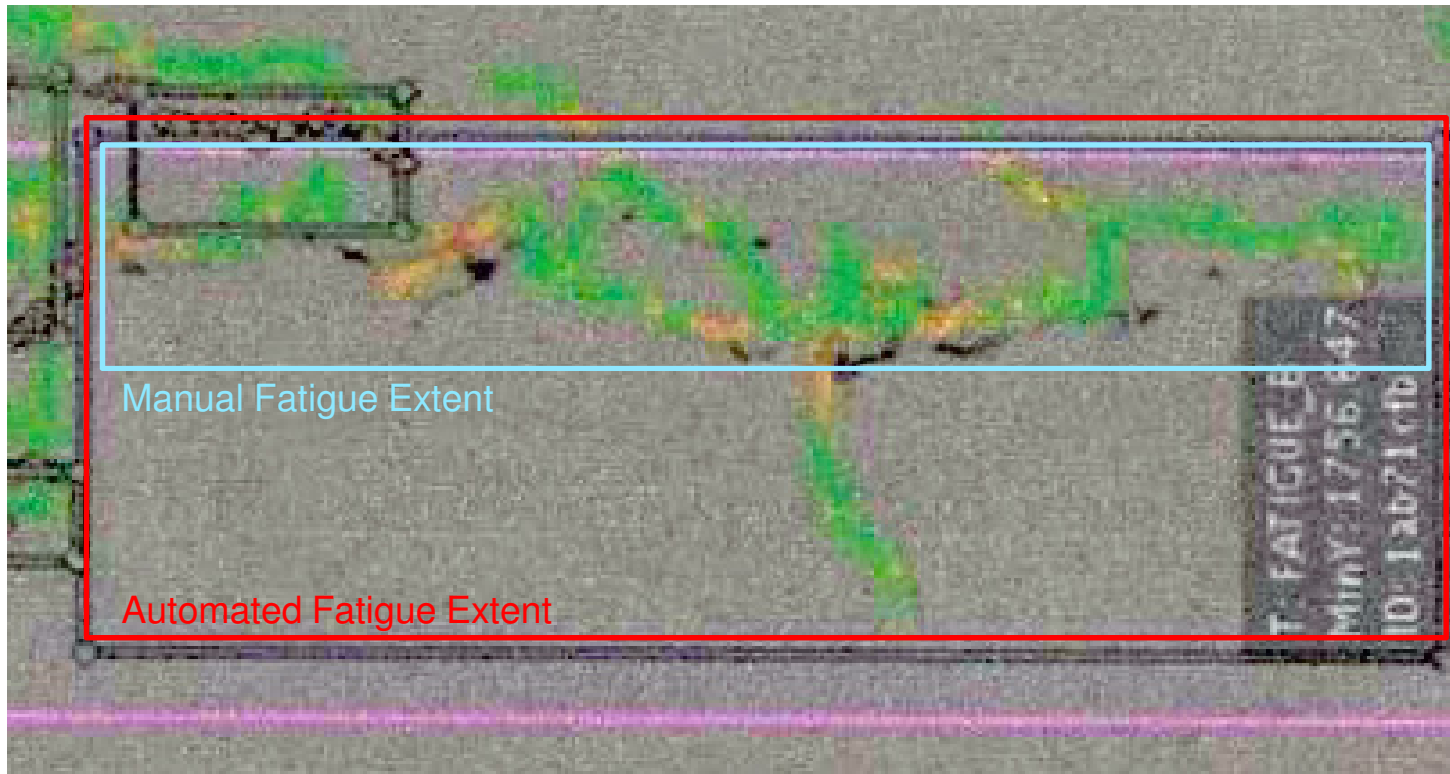


# Distress Characterization & Image Analysis



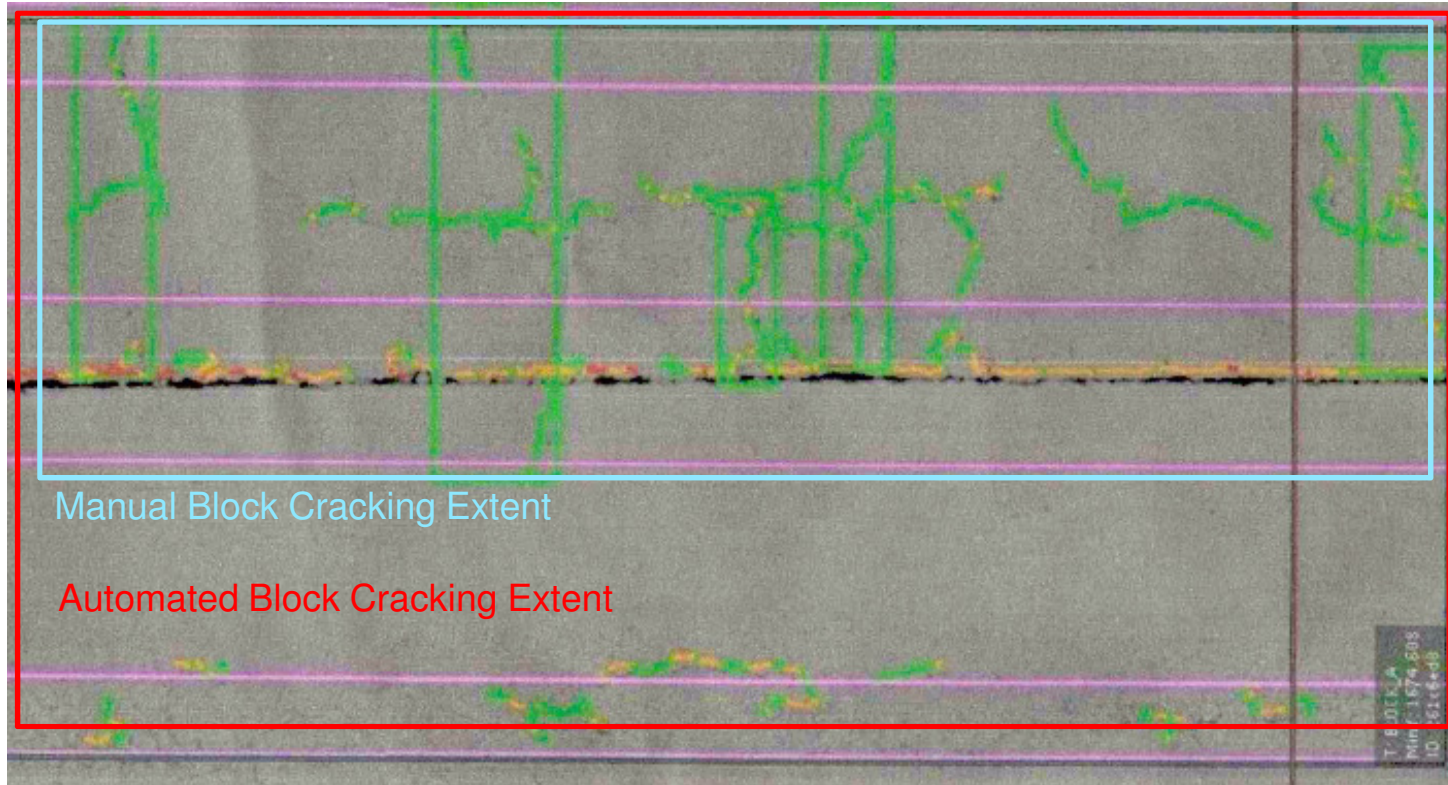
- Training and re-training of AI algorithms is critical to accurate distress detection and characterization.

# Distress Characterization & Image Analysis



Minimize differences between automated and manual measurements.

# Distress Characterization & Image Analysis



Minimize differences between automated and manual measurements.

# Calibration Iterations

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- Adjust wheelpath width in accordance with AASHTO protocols
- Correct lane areas for vehicle deviations
- Adjusted limits for block cracking, transverse cracking, and patching
- Characterized full range of distress types
- Now consider adjustments to OPC calculation



# OPC Distress Index Structure

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OPC = (Functional Index, Structural Index, Non-Structural Index)

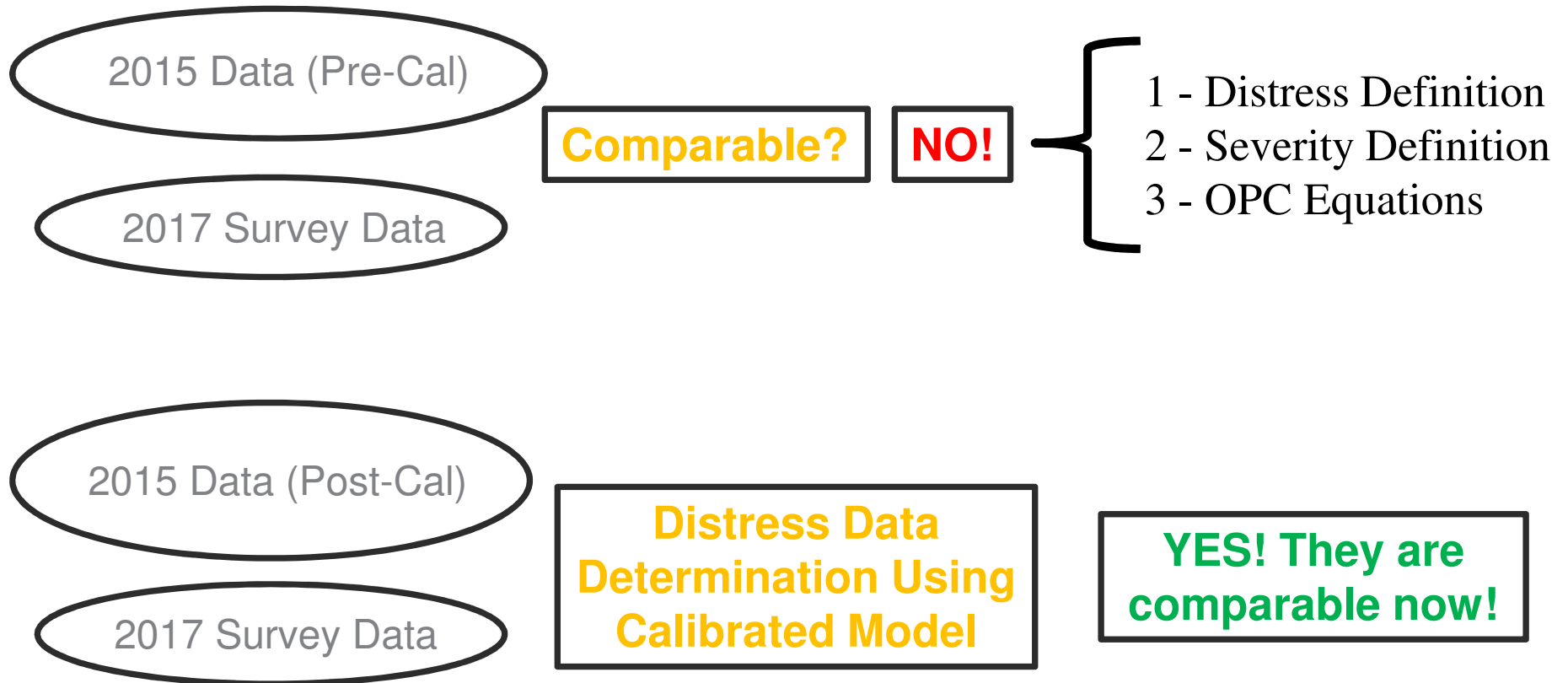


The right treatment?

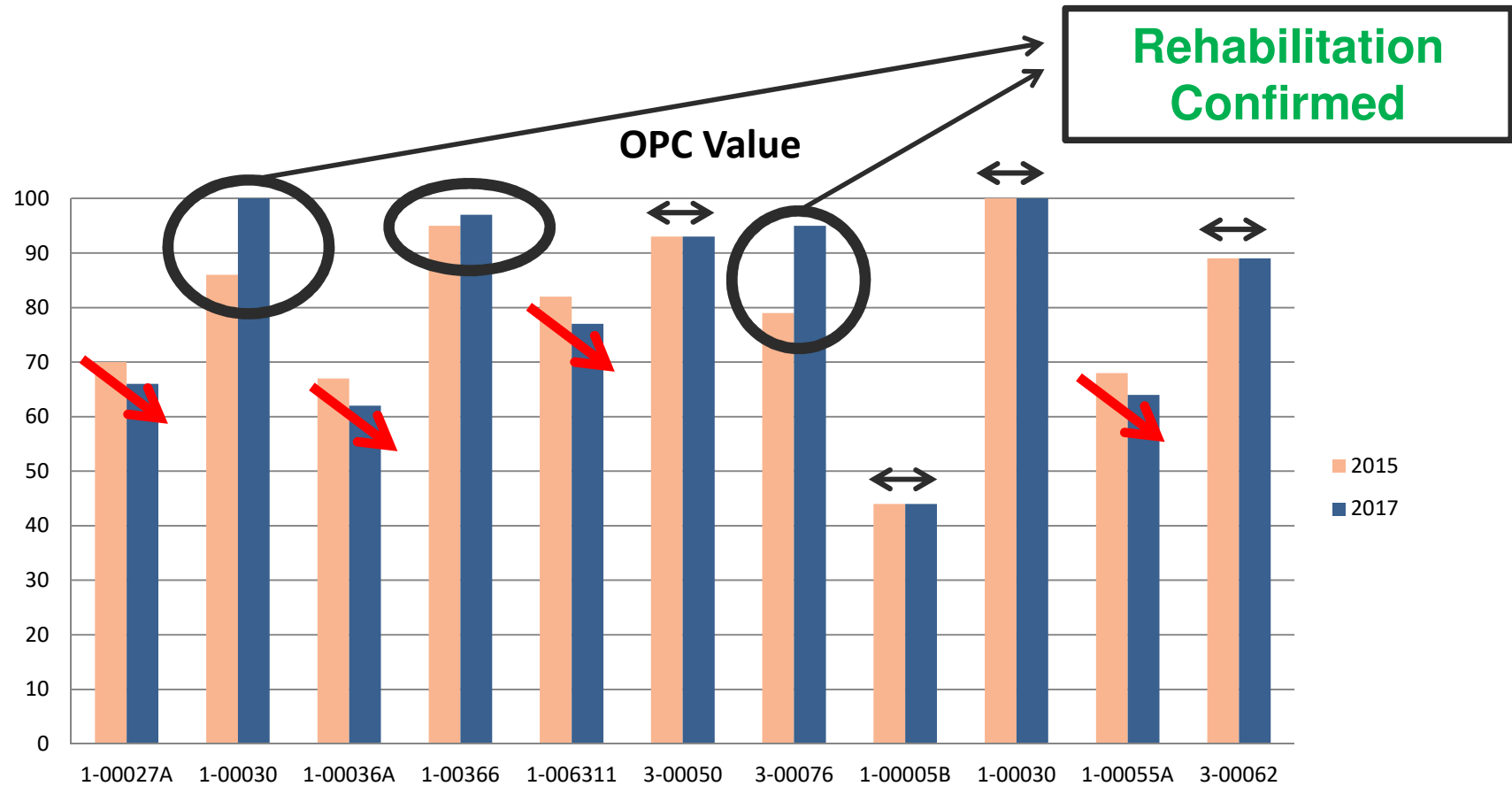


Yes = Stop  
No = Iterate

# Data Comparison



# Historical Data Check of OPC Values 2015 Survey Post-Calibration vs. 2017 Survey



# Key Findings

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- **Raveling:** difficult to characterize using automated methods, reduce impact on OPC scores
- **Coarse Texture and Debris:** high macro-texture and roadway debris limits measurement accuracy
- **Crack Characterization:** challenges in differentiating between block cracking and linear cracking combinations

# Key Findings

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- **Wheel Path Location:** expanding WP width to incorporate more fatigue
- **Limited Distress Types:** inclusion of all distress types in OPC calculation
- **Technology Enhancements:** raveling, ASR, and joint seal deterioration still pose challenges in automated detection



# Recommendations

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- Expand calibration effort by setting up permanent control sites
- Retrain AI algorithms to reduce differences in raw measurements
- Tweak index models and decision trees
- Establish data acceptance criteria



# Future Trends and Recommendations

# Future Trends

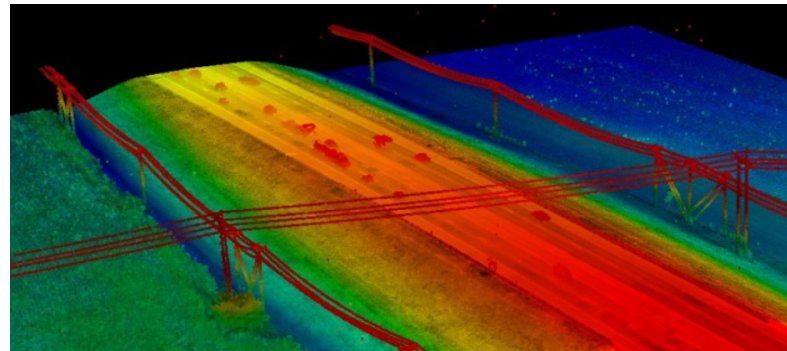
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- AI will continue to advance
- Data collection equipment and methods will change
- Cost to collect and analyze should go down
- Return to the beginning again in our thinking....

# Future Trends

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- AI will continue to advance
  - Improvements to raveling, texture characterization
  - Improvements in crack characterization
- Data collection equipment and methods will change





# Future Trends

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- Cost to collect and analyze should go down
- Return to the beginning again in our thinking (keeping it simple)....



# Recommendations

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- Pavement owner (State, Municipality) should own the protocols for validation, collection, and analysis
- Consider technology advances and vehicle life when purchasing equipment
- Staying current will yield benefits:
  - Cost savings
  - Precision and accuracy



# Questions & Discussion

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