

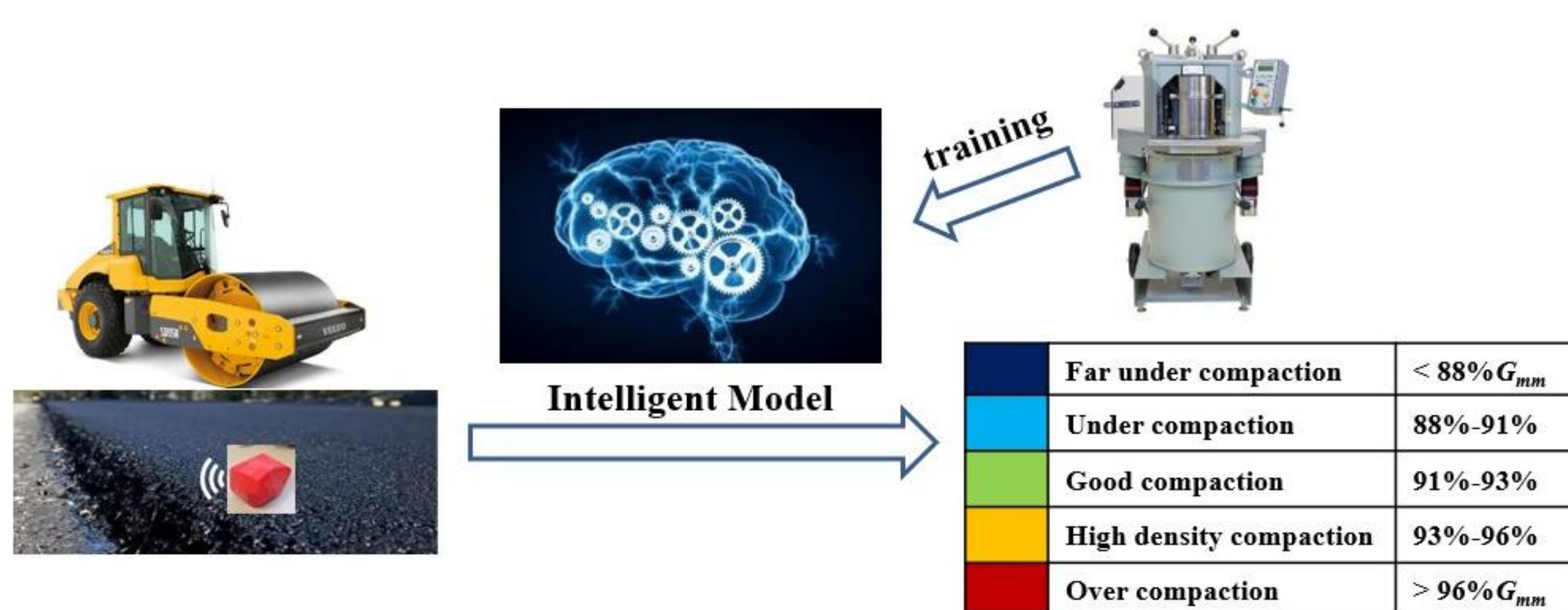
Introduction

- **Compaction** is critical for pavement's performance and service life. Current compaction methods cause uniformity and over/under compaction issues.
- **Intelligent compaction (IC) technology** can improve compaction uniformity but still cannot well monitor pavement compactability.
- **Machine learning (ML) algorithm** and **sensing technology** will be combined for quality control of asphalt pavement compaction.

Objective

- Explain the **compaction mechanisms** of asphalt pavement from **mesoscale**;
- Develop an **intelligent model** to predict the compaction of asphalt pavement to achieve an automated assessment of field compaction

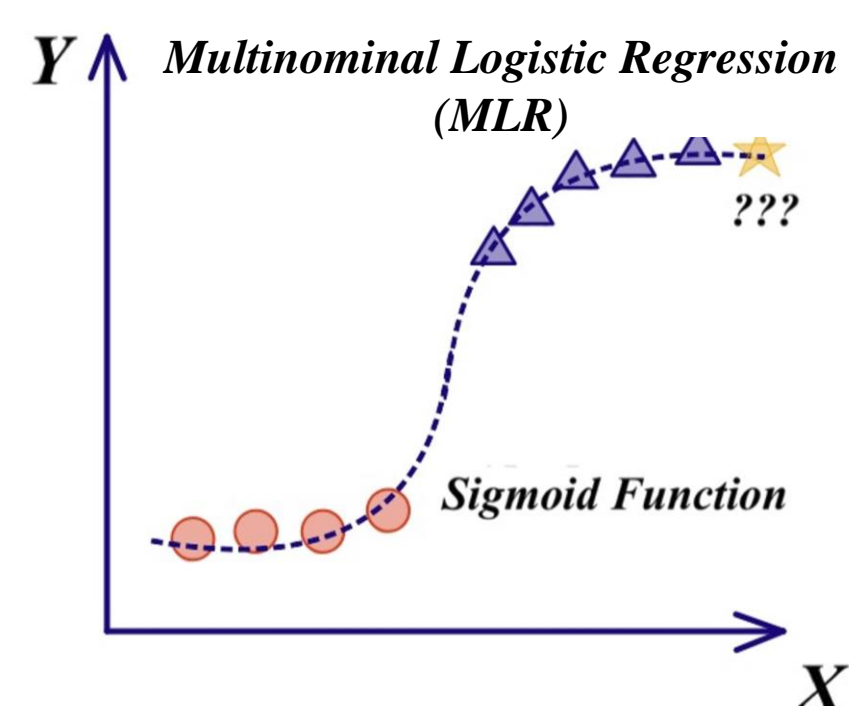
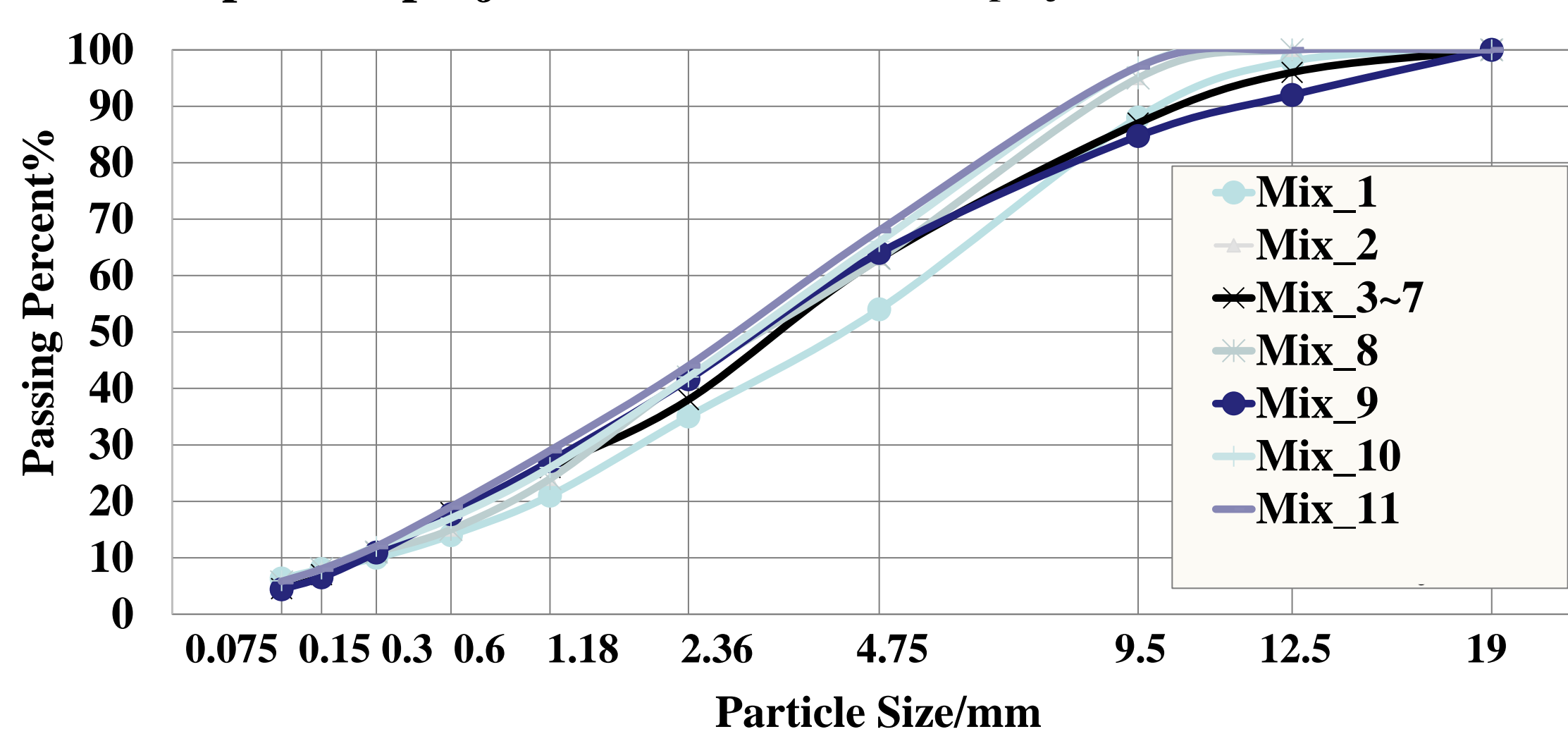
Methodology



- Perception** (Data collection) **Analysis** (Training the intelligent model by lab data) **Decision-making** (compaction prediction)
- The **wireless sensor** and the **ML algorithm** are combined to predict the compaction condition of the asphalt mixture/pavement.
 - **Multinomial Logistic Regression (MLR)** is selected for the compaction prediction **Artificial Neural Network (ANN)** is selected for the density prediction.

Experiment and Model

- **11 asphalt mixtures, total 32 specimens** are compacted in the lab.
- **2 field compaction projects** (Altoona and Indiana project) are carried out.



1	Far under compaction	< 88% G_{mm}
2	Under compaction	88%-91%
3	Good compaction	91%-93%
4	High density compaction	93%-96%
5	Over compaction	> 96% G_{mm}

Input for the ML model

- Particle relative rotation;
- Mixture design information;
- Accumulated compaction energy.

Output for the ML model

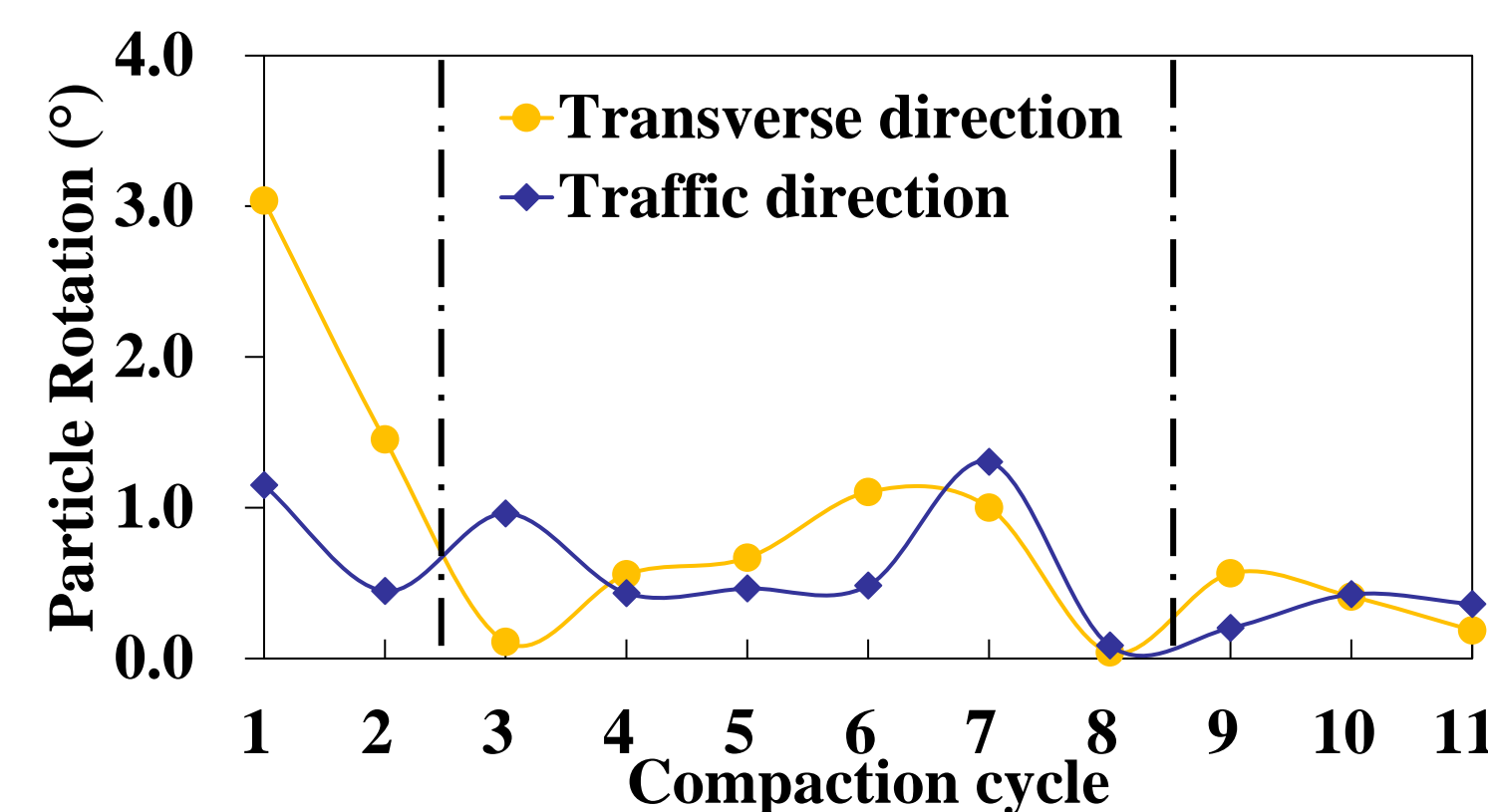
- **MLR**: Compaction condition (5 category)
- **ANN**: Pavement density (% G_{mm})

Results and Analysis

Correlation between the lab and field compaction

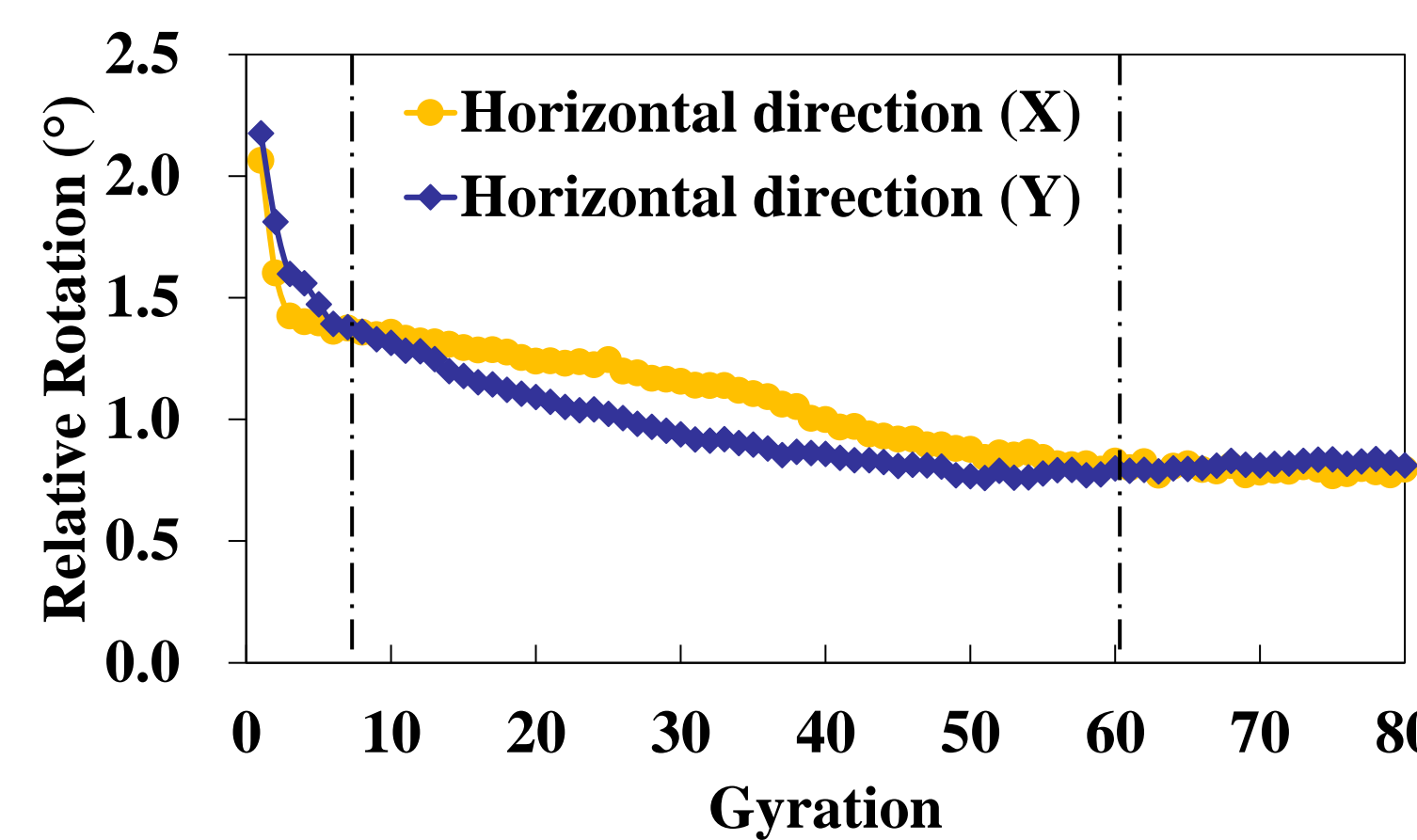
Three-stage patterns are identified in both lab and field compaction:

- **Breakdown stage**: Short, the most dramatic rotation and speedy decrease.
- **Main compaction stage**: Imbalance interaction between compaction loadings and particle shearing resistances.
- **Finishing stage**: Interactions are balanced and particle rotation is static.



Field Roller Compaction

Pearson correlation coefficient:
 $r = 0.806$, meaning:
a **high correlation** between gyration and field compaction.



Lab Gyration Compaction

$$r = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum (x_i - \bar{x})^2 \sum (y_i - \bar{y})^2}}$$

Prediction Results

- Two **field projects** can be reasonably predicted based on the **Machine Learning (ML) model training by the laboratory compaction data**.

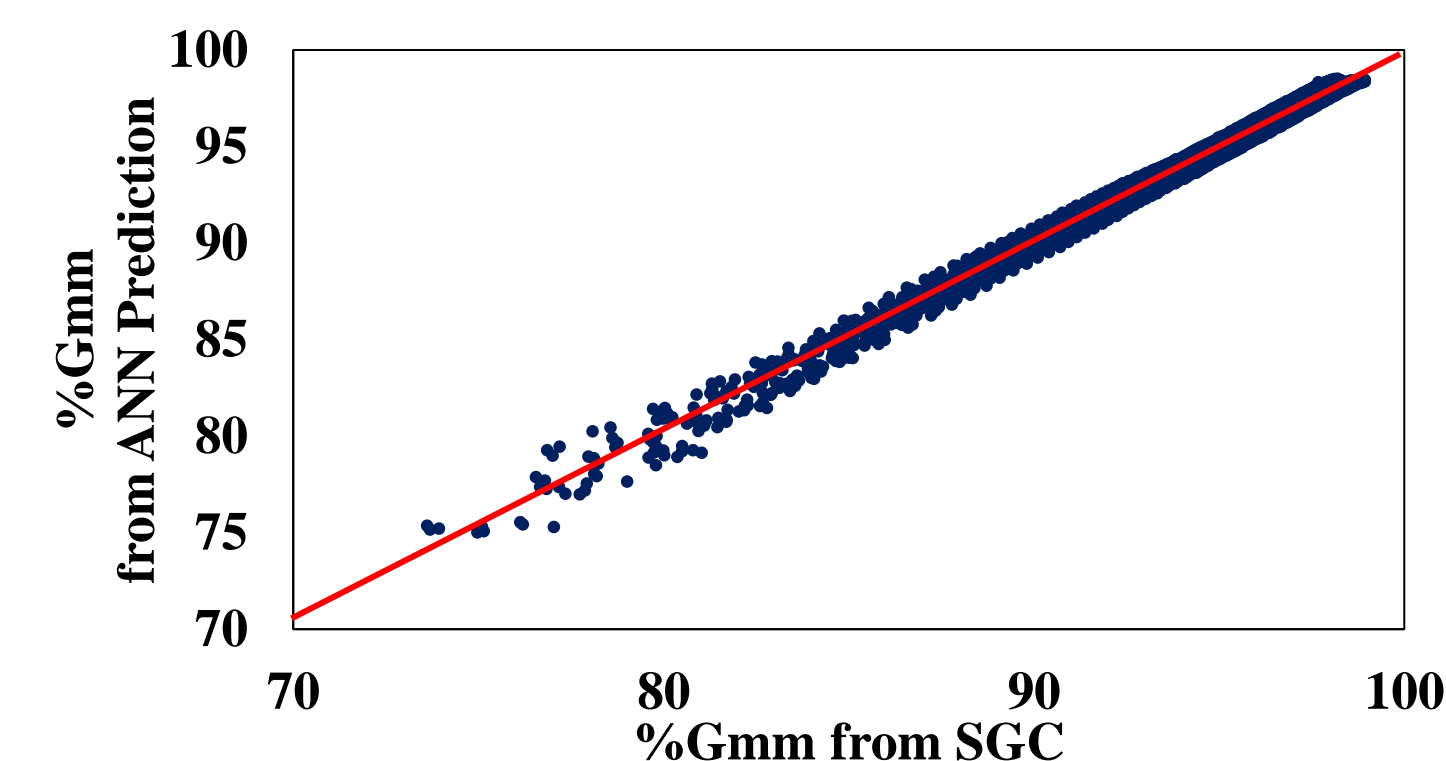
Laboratory Compaction Prediction:

MLR model :

- 90% Accuracy

ANN model:

- Mean Relative Error (MRE) < 0.5%



Project	Altoona Project											
	Cycle	1	2	3	4	5	6	7	8	9	10	11
Roller	V	V	V	S	S	S	V	V	V	S	S	
MLR	1	1	1	1	2	2	2	3	3	3	3	
ANN	78.2	83.7	88.3	88.8	89.8	90.3	91.4	92.2	92.4	92.8	93.1	
Project	Indiana		Note:									
Cycle	1	2	➤ The Class 1,2,3 means the Far under compaction, Under compaction and Good compaction, respectively;									
Roller	V	V	➤ Core density is the density of the core sample (% G_{mm}) as the true value for verification.									
MLR	2	3										
ANN	91.7	93.1	Core density: 92.8; 92.5									

Conclusions

- **Three-stage compaction patterns** are identified for both lab and field compaction based on **particle rotation**.
- **Multinomial logistic regression (MLR)** and **Artificial Neural Network (ANN)** are appropriate to predict the compaction conditions of the asphalt pavement.
- A reasonable density can be predicted by the ML model trained by the lab data using the **particle rotation, compaction energy, and mixture design** as inputs.