



IRSO
INTERNATIONAL RAIL SAFETY
ORGANISATION



SC 2023

INTERNATIONAL RAILWAY
SAFETY COUNCIL

"Reshaping Railways in an Uncertain World"

CAPE TOWN, OCTOBER 1 - 6, 2023

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HOSTED BY  RAILWAY SAFETY
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REGULATOR
RAIL SAFETY ON THE RIGHT TRACK

Dr. Anjana P Das
TATA ELSXI Ltd, India

Multi-component detection in Rail Track joint bar using YOLOv5



India's Most Valuable Brand

India's Only Value-based Corporation – a Visionary, a Pioneer, a Leader, since 1868

9,35,000+

Employees

\$ 128 B+

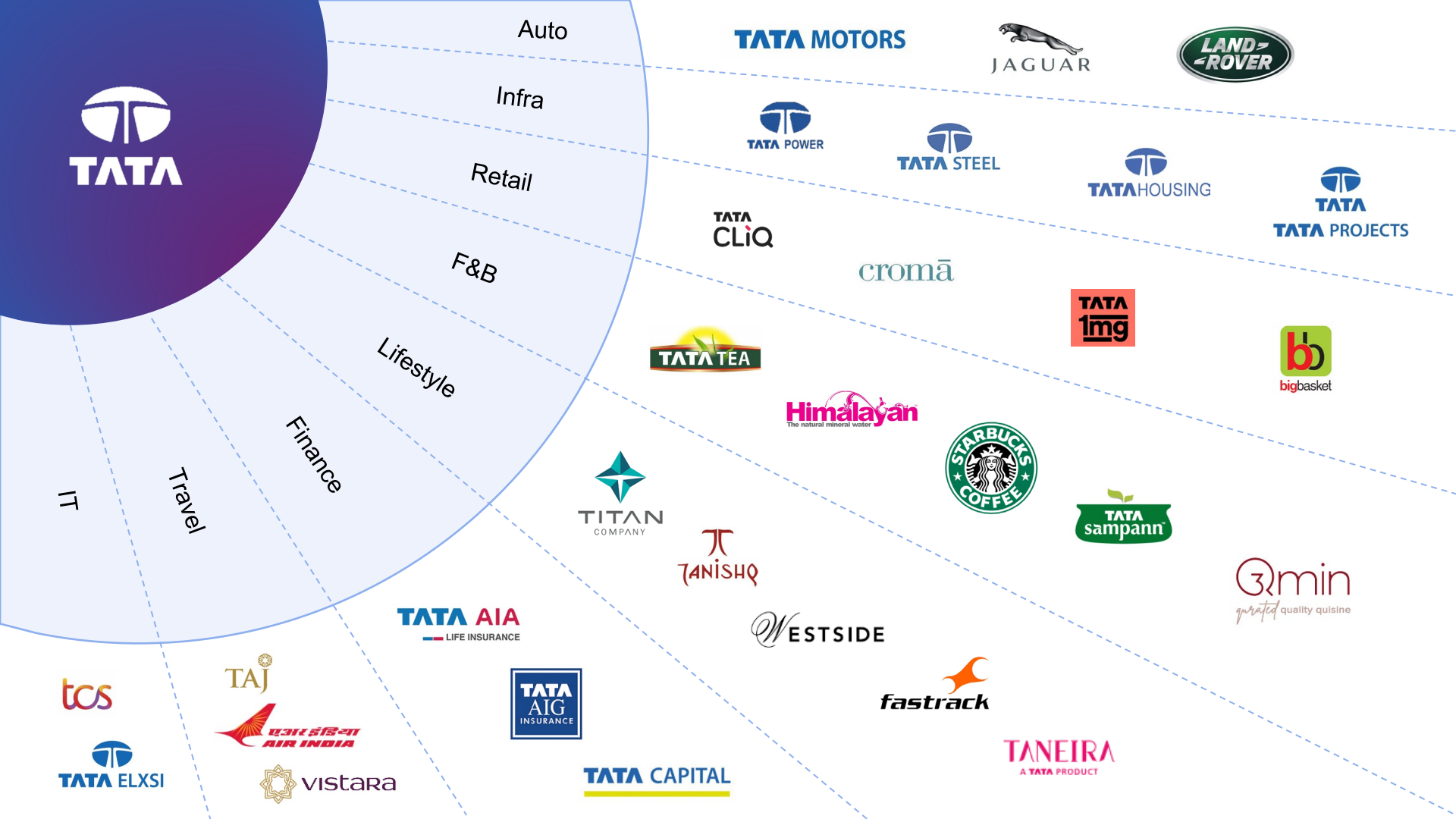
Revenue

150+

Countries

150+

Years of Legacy



TATA ELXSI BUSINESS OVERVIEW

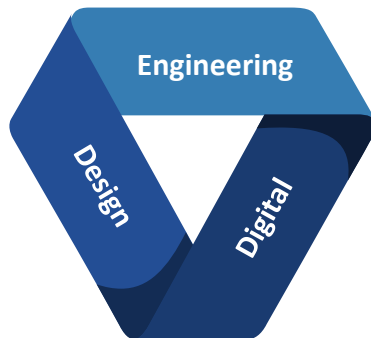
12000+
Engineers

13.4%
YoY Growth

32+
Years in
Business

33
Cities
Worldwide

CORE SERVICES



Transportation

RAIL

- Smart Mobility
- Rolling Stock and Systems
- Wayside & Signalling



AUTOMOTIVE | OFF HIGHWAY

- Passenger Experience
- Connected & Autonomous
- Shared & Electric



Media & Communication

BROADCAST & MEDIA

- OTT Streaming
- RDK, Android TV, CPE
- QoE, QoS, Customer Experience



COMMUNICATIONS

- 5G, SDWAN
- Network Transformation
- Digital Transformation



Healthcare

MEDICAL DEVICES

- Product Design
- Systems Engineering
- Regulatory Compliance



PHARMACEUTICAL

- S** Safety
- Packaging & Labelling
- Pharmacovigilance



Established
1989



Global HQ
Bangalore

Global Presence

Headquartered in **Bengaluru** with offices around the globe including NA, EUROPE & APAC

NORTH AMERICA

Atlanta, Canada, Irvine,
Naperville, Philadelphia, Santa
Clara, Troy, Toronto

EMEA

Dubai, France, Germany, Ireland,
London Digital Studio, Netherlands,
Poland, Portugal, South Africa, Spain,
UK

APAC

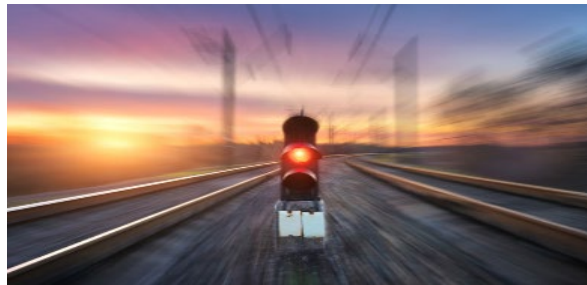
Japan, Malaysia

INDIA

Bengaluru (HQ), Chennai, Hyderabad,
Mumbai, Pune, Trivandrum



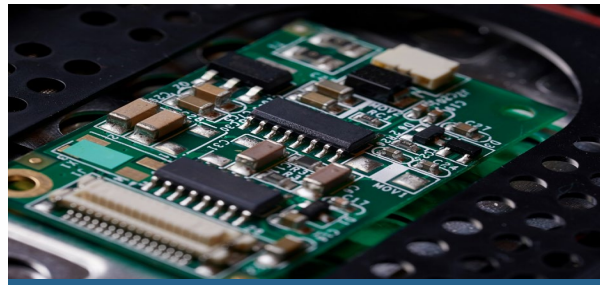
Our Focus In Rail industry



Asset inspection and operations – **Safe and Efficient Operations**



Object detection, Perception systems
(**Autonomous Systems**)



Safety Systems Development
(**Full Ownership**)



Crunching the Vehicle product development cycle – Quick time to market
(**Competitive Advantage**)



Seamless passenger experience, Passenger security & Surveillance (**Smart Mobility**)



IoT/Cloud for asset management and preventive maintenance (**Rolling stock & wayside**)

Inspection scope areas in Railroad operations



Track Inspection

Geometrical

- Curvature
- Gaps in assemblies
- Measurements

Structural

- Missing, loose parts

Damages

- Rail end batter
- Cracks on rail
- Broken, damaged crosstie



Rolling stock

Geometrical

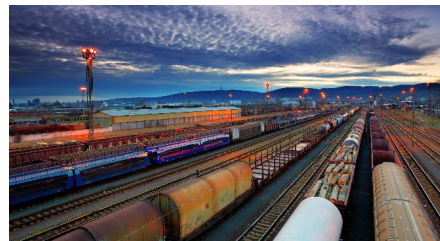
- Couplers
- Structural deformation

Structural

- Door/ lid status
- Container contents

Identification

- Engine/ wagon ID



Yard

Structural

- Signage Health

Surveillance

- Theft
- Cleanliness, spillage



Way-side

Structural

- Signage Health
- Overhead Vegetation

Civil Structures

- Bridges
- Tunnel

Environment

- Landslide
- Object on track

Significance of Track Inspection



- Safety
- Maintenance Planning
- Legal and Regulatory Compliance



- Asset management
- Performance monitoring
- Continuous Improvement



- Early Detection of Issues
- Operational efficiency
- Risk mitigation



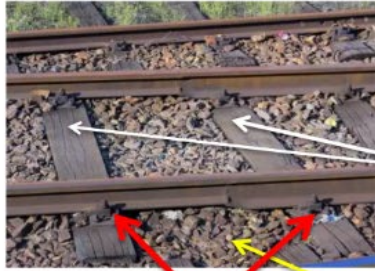
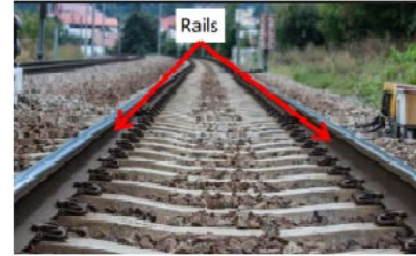
- Customer satisfaction
- Enhanced Reliability
- Financial impact
- Reputation and Public Perception



Rail Track Elements



Fish plates / Joint bar



Fastenings

Sleepers



Ballast

Rail Joint Bar

Joint bars are an essential component of rail infrastructure used to connect two distinct rail tracks without the need for welding.



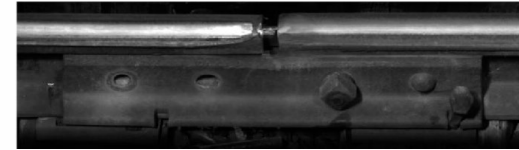
Bolt head



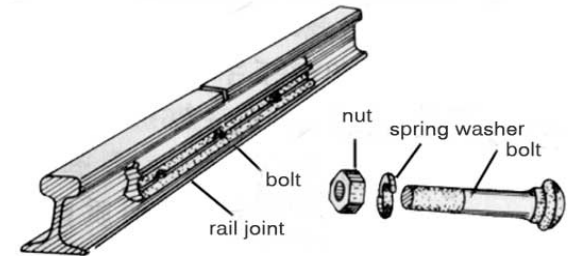
Nut



Holes

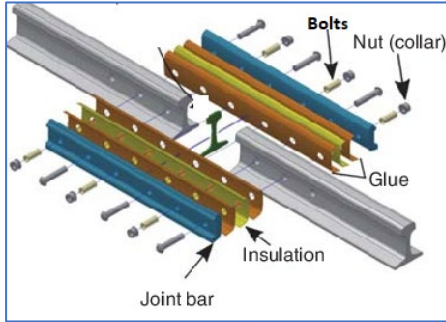


Joint Bar with two Missing Bolts Captured by the System

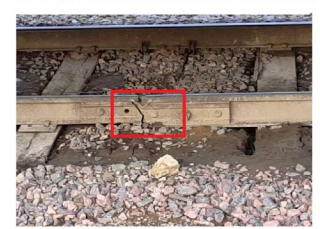
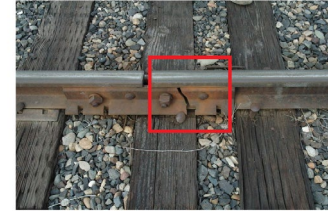


Defects in joint Bars

Joint Bar Integrity

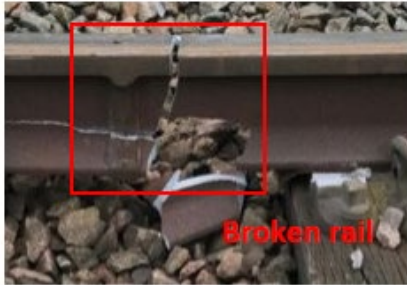


Few versions of missing bolt



Few versions of crack

Rail Breakage



Missing bar



Rail fastener Defects



Conventional Approaches

Manual Inspection



Shortcomings of Manual Inspection

- Human Error
- Disruptions to the smooth functioning of railways.
- Time-consuming and Labor-intensive
- Limited geographic coverage

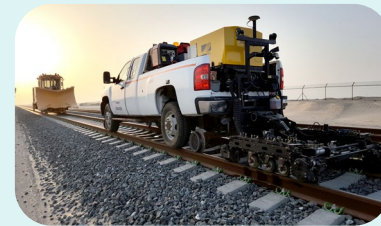
Shortcomings of Inspection Vehicles

- High Capital Cost
- Limited Accessibility
- Operator Dependency

Significance of Image-Based Solutions

- Research gap in image-based solutions
- Lack of literature work on defect detection , gap measurement from camera images

Inspection Vehicles



Joint Bar inspection



Track gage Inspection

Inspection with Drone Technology



Reliability in processing the experimental field data and defects



Localization of defect in a particular area using latitude and longitude



GUI based approach for method and results



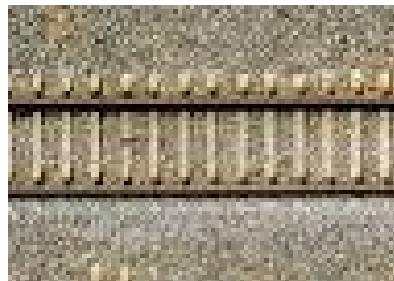
High-quality images that contain large information for monitoring and analysis

Inspection Solution is Executed in Edge Computer



Cascaded Approach: Step-by-step detection of target component to eliminate errors

Hybrid of Computer Vision & AI/ML algorithms to achieve high accuracies

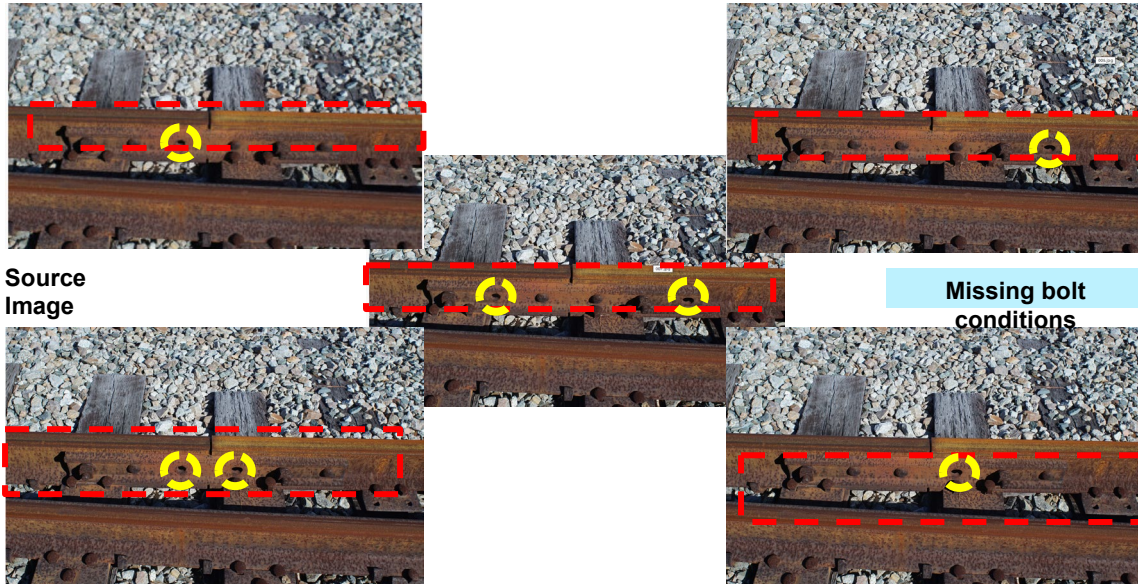


Development Approach – Synthetic Data Generation

Best Practices/ Innovations

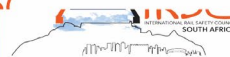
Synthetic Data Generation tool – **Able to generate millions of training data from limited parent set**

Customized Generative AI models

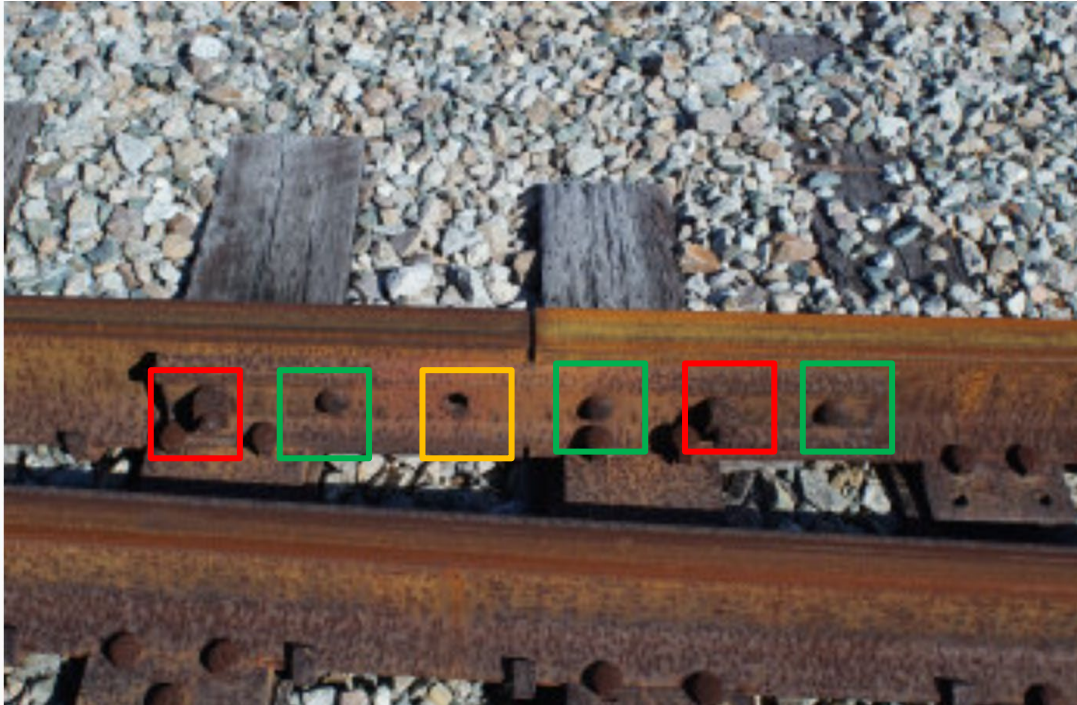



Automated 3D Model Generation

Loose bolt conditions




Multi-Component Detection



Nut_and_Bolt 

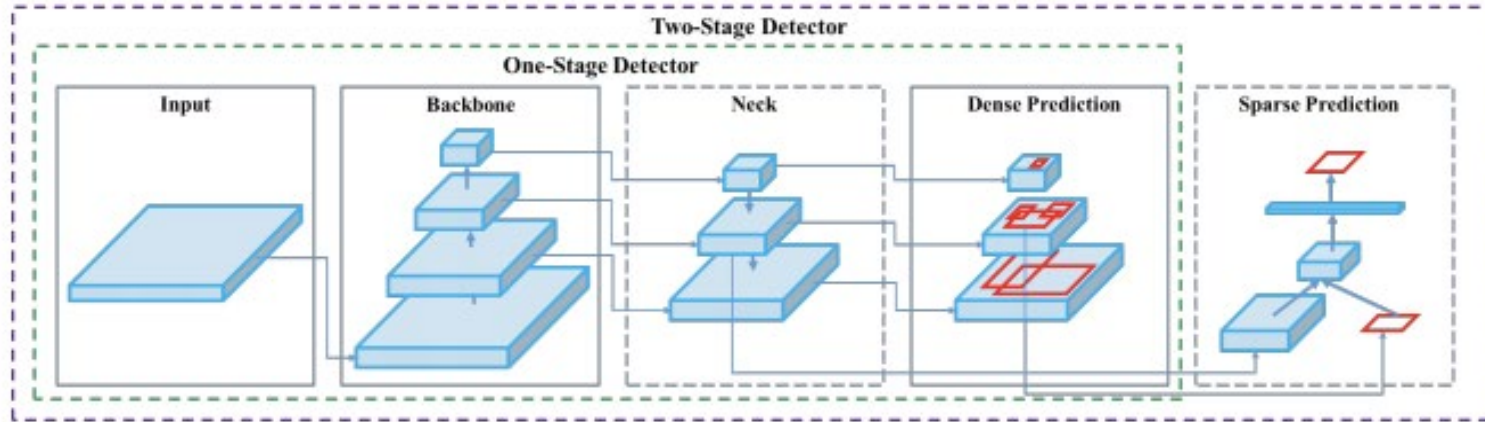
Bolt_head 

Missing_nut
_and_bolt 

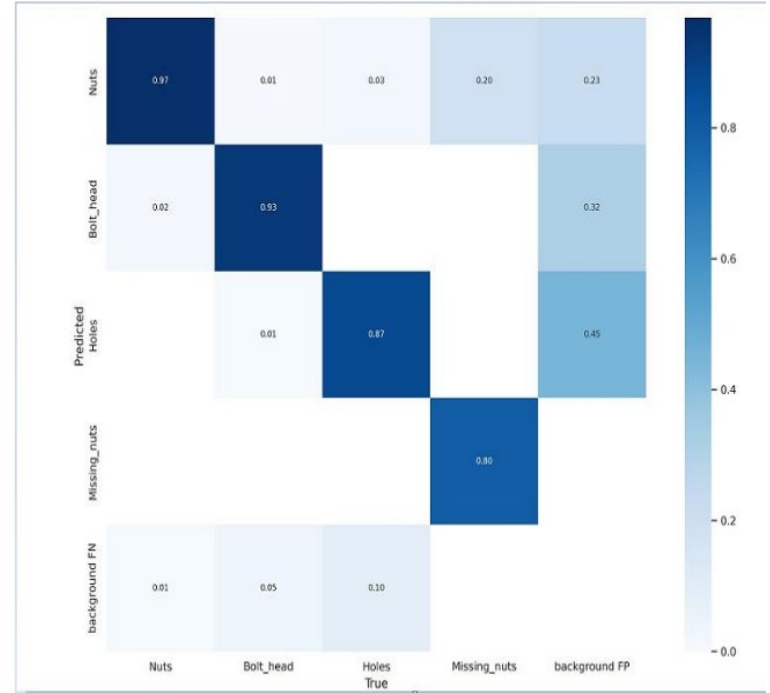
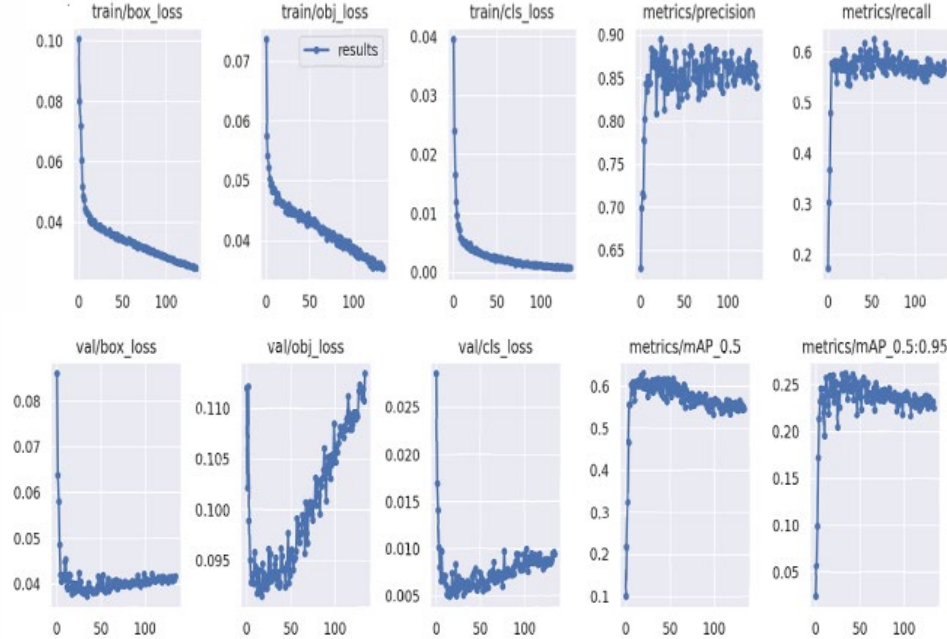
Multi-Component Detection-YOLO V5

The YOLO network consists of three main pieces.

1. **Backbone**: A convolutional neural network that aggregates and forms image features at different granularities.
2. **Neck**: A series of layers to mix and combine image features to pass them forward to prediction.
3. **Head**: Consumes features from the neck and takes box and class prediction steps.



Experimental Results and Inferences



The training procedure took an average of 10 seconds for 86 batches during one epoch, and the evaluation process took an average of 9 seconds for 22 batches. Total execution time was 0.435 hours i.e. 26.1 minute for 100 epochs.

Experimental Results and Inferences

Hyper parameters	Values used for training	Model	Training Results -Prediction (%)
Learning Rate [lr-Final one cycle LR] Momentum Classification Loss gain Object loss gain	0.0001 0.984 0.5 1.0	Yolo5s.pt	Nut - 97 Bolt head - 93 Holes - 87 Missing Nut - 80
Learning Rate [lrf-Final one cycle LR] Momentum	0.001 0.937	Yolo5s.pt	Nut - 96 Bolt head - 89 Holes - 82 Missing Nut - 80
Learning Rate [lr] Learning Rate [lrf-Final one cycle LR] Classification Loss gain Object loss gain	0.01 0.01 0.3 0.7	Yolo5s.pt	Nut - 98 Bolt head - 83 Holes - 89 Missing Nut - null

Class name	Precision value	Recall value	Accuracy (%)
Nut	82.9	98.9	97.0
Bolt_head	80.1	94..8	94.6
Holes	73.1	89.6	87.6
Missing nut	0.64	98.7	80.0

Hyper parameter tuning constructed based on the learning rate and momentum and further optimized using the Adaptive Moment Estimation (ADAM) optimizer



Conclusion



IMPACT:

50x times savings in time for inspecting the same length of track compared to manual method.

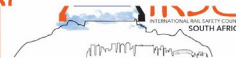
KEY DIFFERENTIATORS

AUTONOMOUS

NON-INVASIVE

HIGH FIDELITY

NEXT GEN RAIL ASSET INSPECTION SOLUTION OPENS UP NEW BUSINESS MODELS





Thank You

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