



Knowledge Transfer **Partnerships**



Drone derived Multi-Modal Al Predictive Model to detect & identify peatland degradation, plan restoration activities & monitor peatland health











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Background

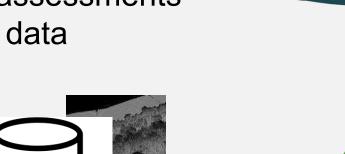
Healthy Peatland Vs Degraded Peatland

Benefits of healthy peatlands Impacts of degraded peatlands

The Challenges with the Current Approach

Manual Approach

Based on ground based assessments or low resolution satellite data



1. Ecologists walking the peatlands & collecting data (mapping erosion feature)



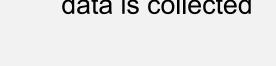
2. Unstructured/incomplete data is collected

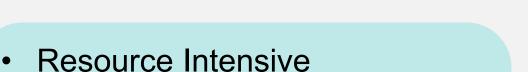


3. Approximate

land measurement

& Plan Restoration





Time-consuming

 Subjective observation Lack precision & efficiency

 Hinder effective restoration planning & decision making

Expensive to deliver, lack of scalability, limited reduction of carbon emissions

Our Approach: Multi-Modal Al Predictive Model

Windowed

Tiles

RGB

Drone & IoT Devices

Scenario

UK has 3 million ha of peatland which

• 80% in poor condition & releases 23

Our AI based automated tool will enable

restoration activity cost effectively &

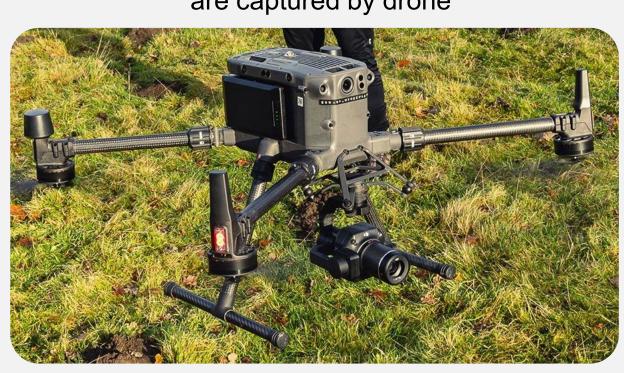
reduce carbon emissions faster

million tonnes of CO₂ annually ¹

scalable planning of peatland

stores over 3 billion tonnes of carbon 1

High resolution RGB & LiDAR images are captured by drone



Water Table depth (WTD) & CO₂ Concentration are collected from remotely accessed IoT sensors



Gas Analyser



Store



Peat Camera

Multi-Modal Data Drone RGB Image Drone LiDAR image CO₂ Flux

Remote IoT Data

RGB & LiDAR

Preprocess

Filter **Eroded Area**

Framework

Deep learning model

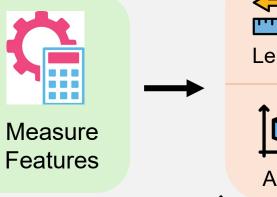




Plan Peatland

Restoration Activity

Post Process



Volume

Delivery of peatland

restoration

Detected Erosion Feature

Automated detection

 Accurate measurement

Scalable planning

Expedite restoration

Business carbon

audit trail

 Cost-effective restoration

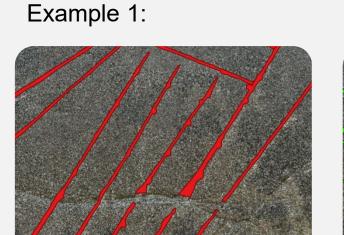
reduction

Reduce Carbon Emission

Outcomes

Map of Erosion Features

Identify areas undergoing degradation

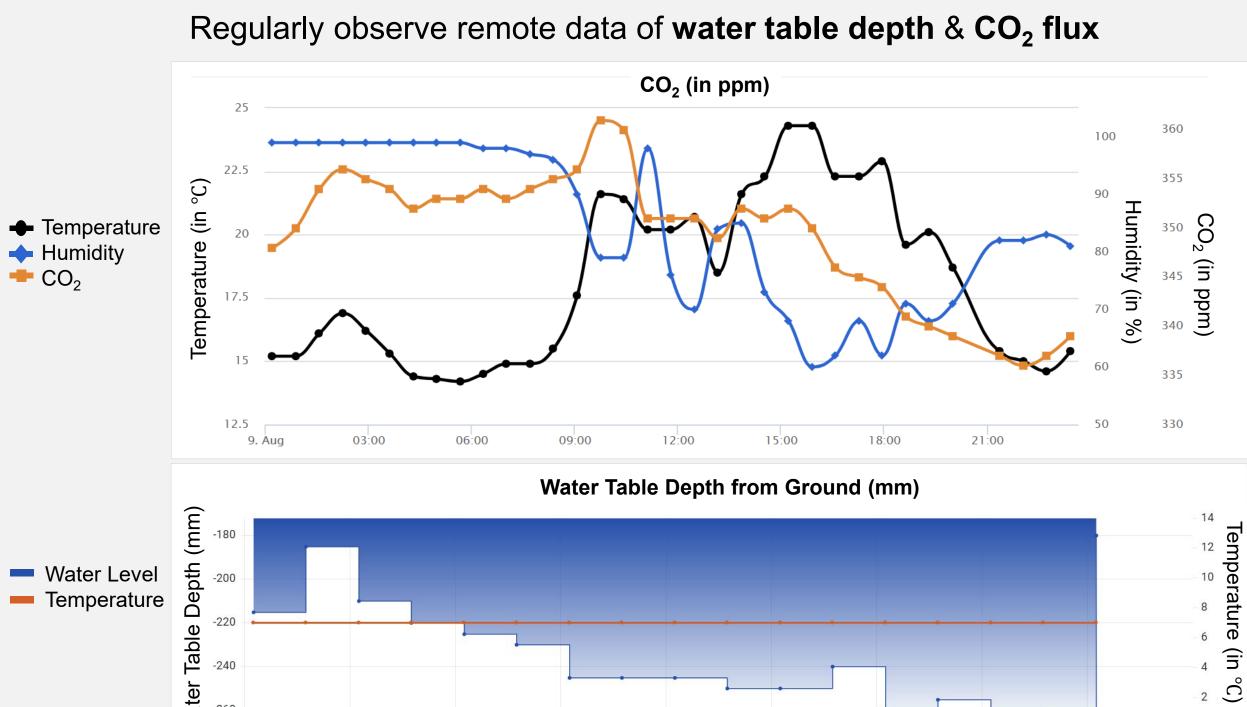




Manual Annotations (Red) vs Model Predictions (Green)

Peatland Health Assessment

Monitor Peatland Health



Restoration Plan

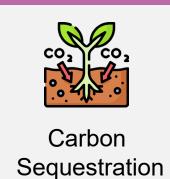
Plan restoration based on erosion features measurement & peatland health assessment

Carbon available: 2000 tC02e (over 20 years) ²

Example: Peatland restoration - Grip Blocking Location: North Yorkshire Restoration size: 50.5 ha



Benefits

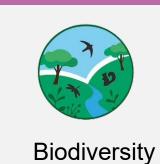


Example 2:







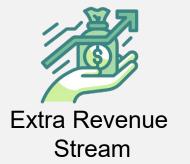




- CO₂









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