

Detecting Flood Embankment Deterioration and Future Projection

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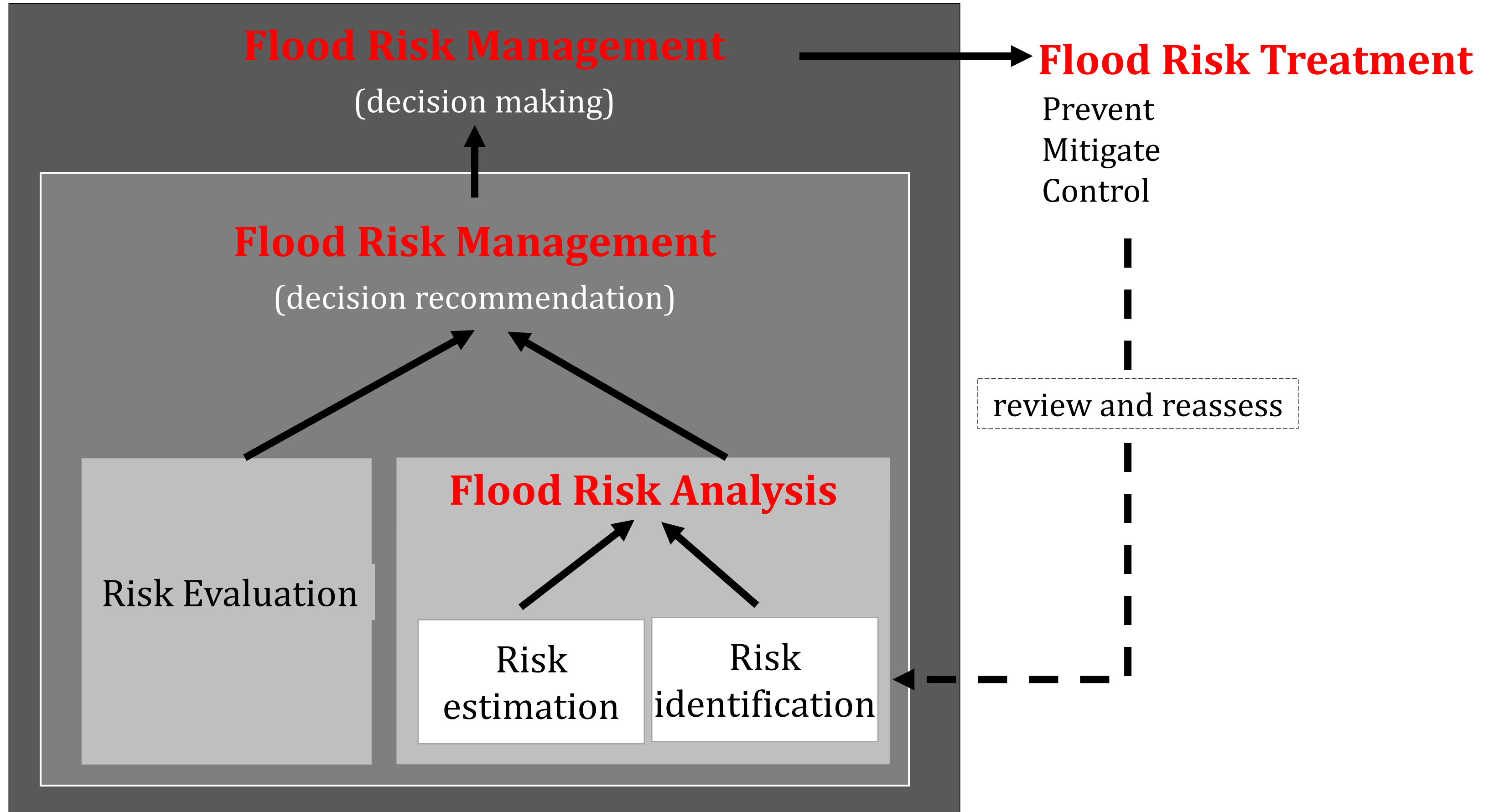
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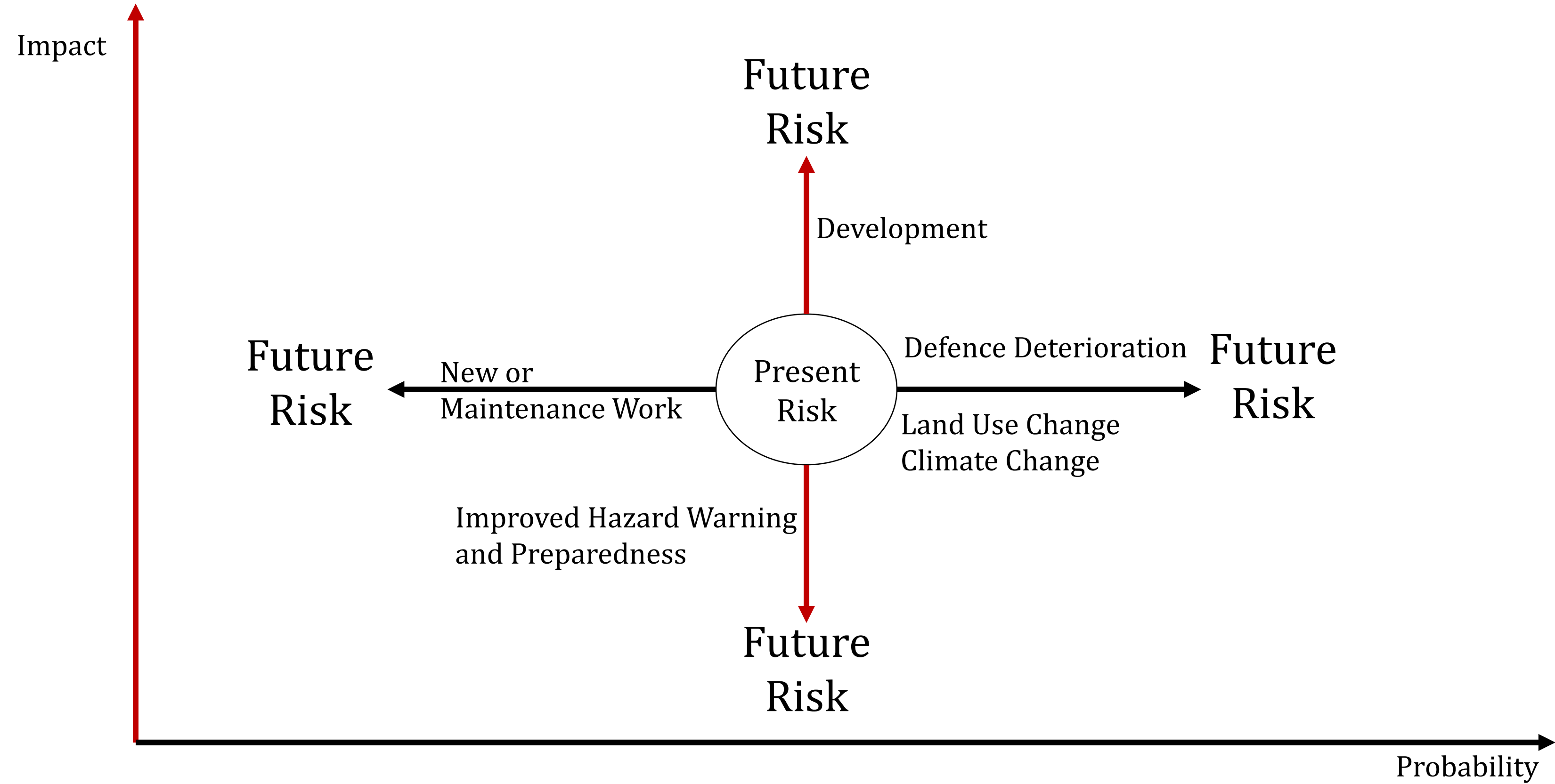


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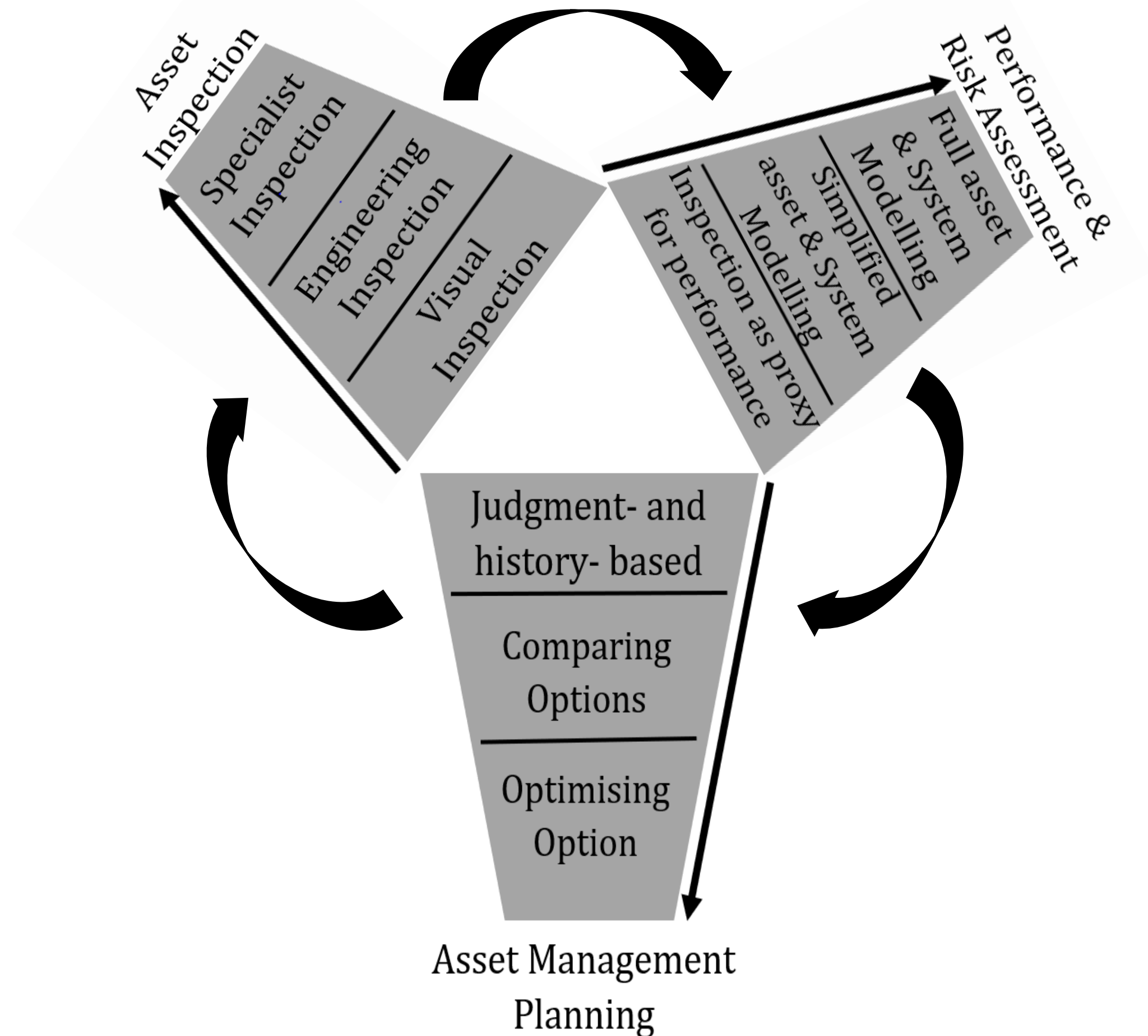
Outline



Outline



Example approach of flood defense asset management



Flood Disaster

Key Point

*“Globally, during 2000 – 2019, **rapid urbanisation** increased potential of disaster susceptibility, and extreme weather events have caused thousands reported disasters, **many of fatalities**, and continue to cause billions of dollars of direct **economic loss**. And under the background of global warming, such losses will continue to increase in the future.”*

Flooding is defined as the temporary presence of surface water, on or near an embankment, and as the most frequent and widespread natural disaster in the world and typically destructive (1).

Flood Data (2000 – 2019) compared to other types of disasters

1 st	Floods are the most common type of disasters, accounted for around 3254 (44%) of total events.	1 st	The type of disasters that affecting 1.6 billion people worldwide or around 41% of total affected population.
3 rd	It accounted around 636 billion US\$ (21%) economic losses.	4 th	One of the deadliest disasters with at around 104.614 fatalities.

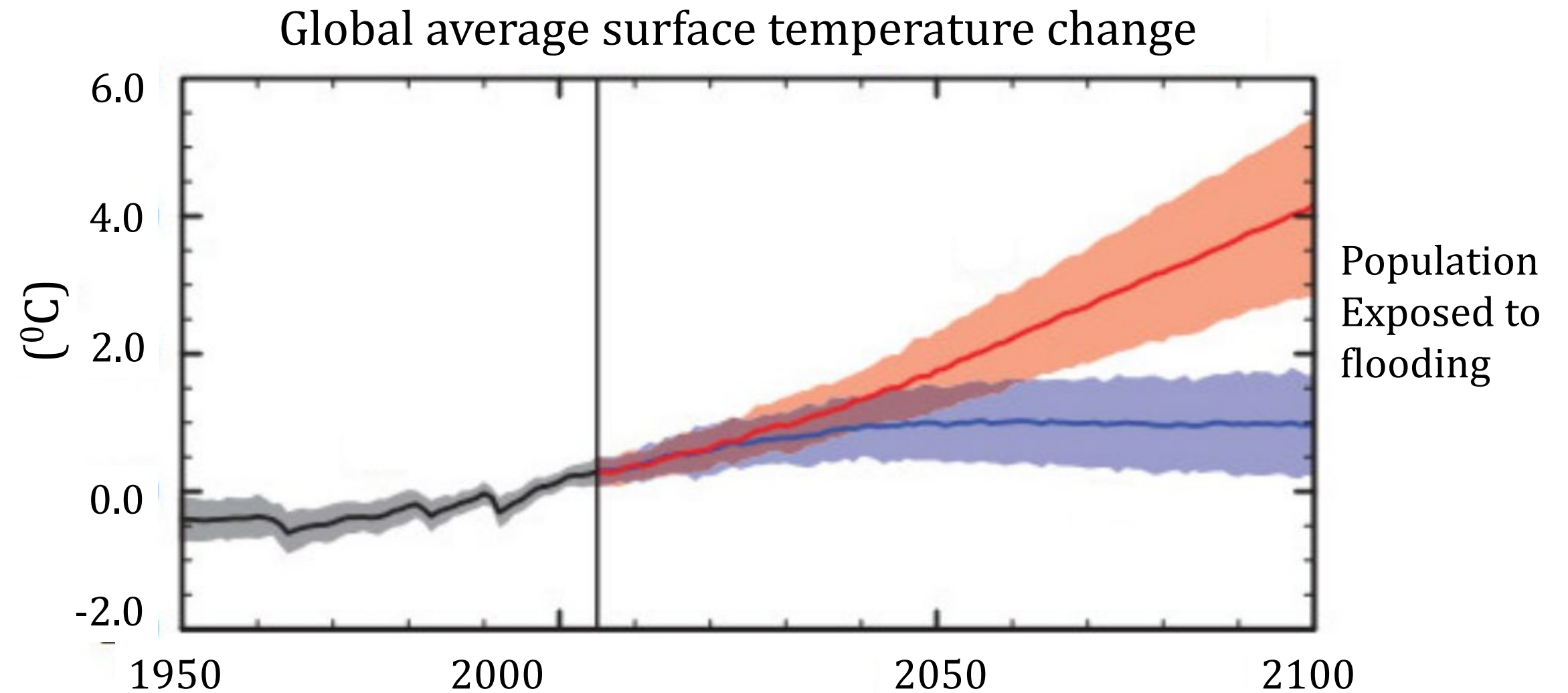
Source:
[1] Iqbal et al. (2021), [2] Centre for Research on the Epidemiology of Disasters (CRED), United Nations Office for Disaster Risk Reduction

Flood under Climate Scenario

Key Point

Better flood control, including prediction and monitoring, is one of the possible solutions in DRR policy terms since affordable and effective technologies already exist.

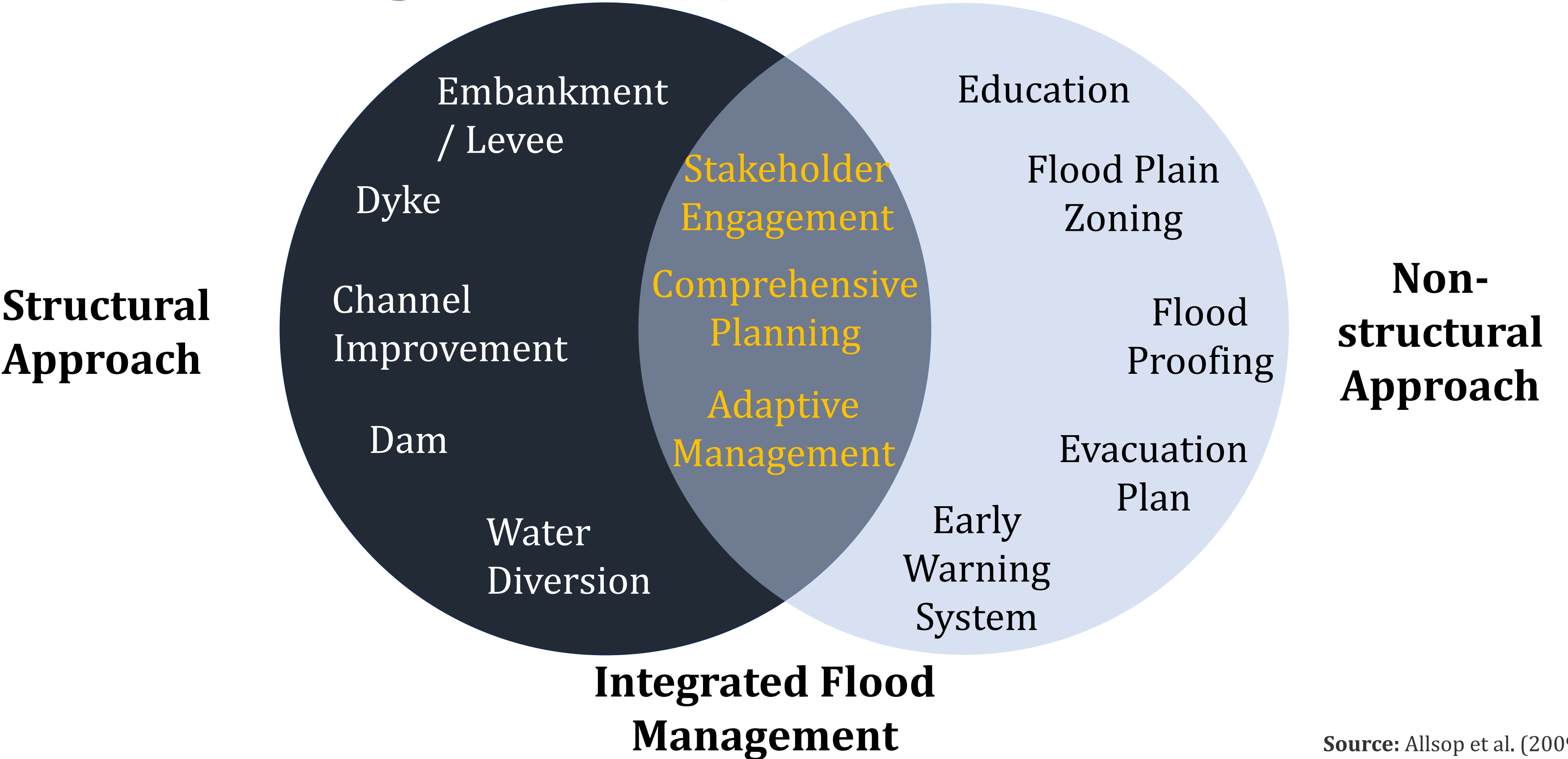
*The priority should be given to **cost-effective measures** in poor regions at high risk of recurrent flooding.*



- People living in **coastal cities and riverine areas** are considered among the most vulnerable to sea level rise, storm surges, and coastal flooding.
- **Global warming** is estimated to increase the frequency of potentially high impact natural hazard events across the world.

Source: Centre for Research on the Epidemiology of Disasters (CRED), United Nations Office for Disaster Risk Reduction

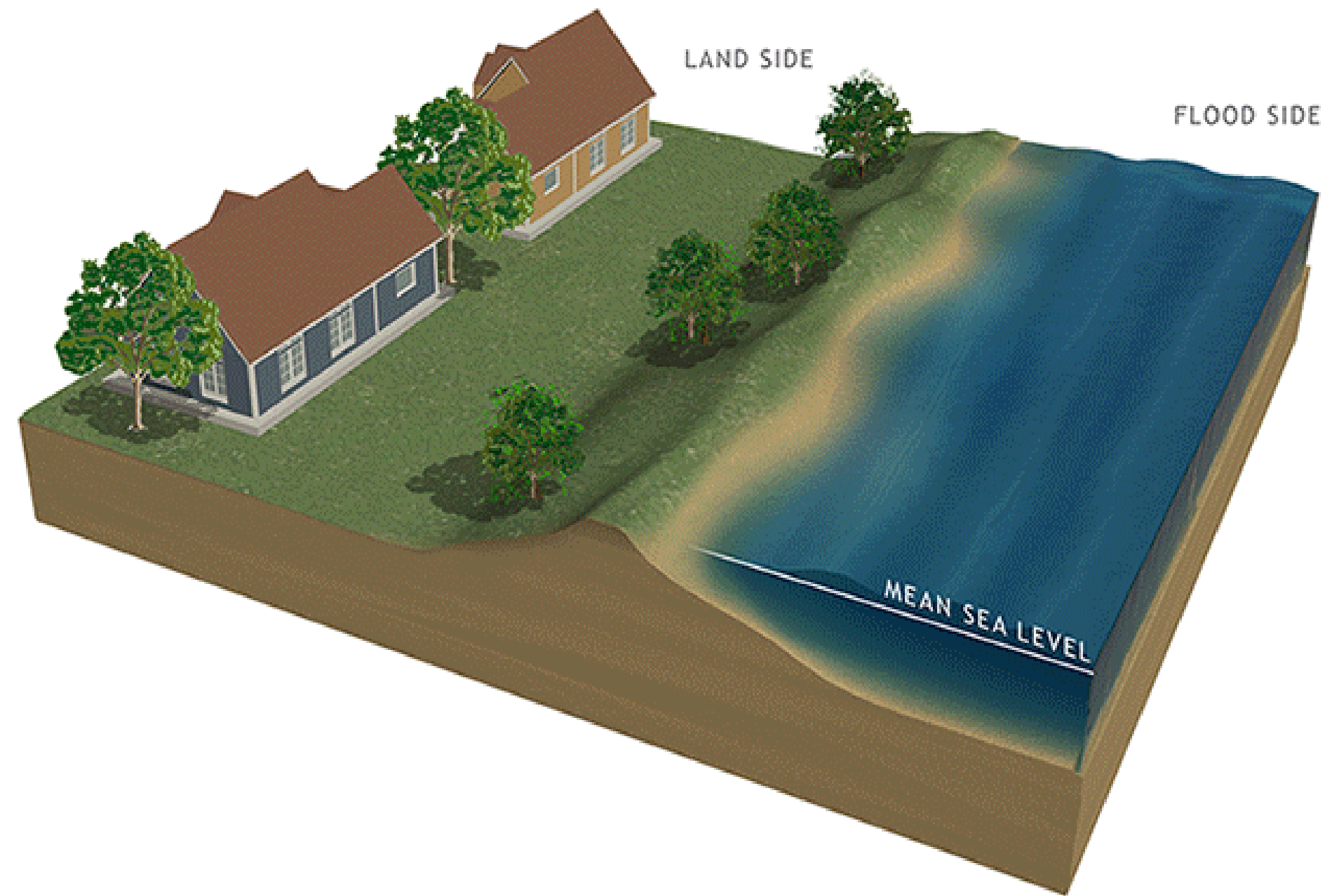
Flood Management Approach



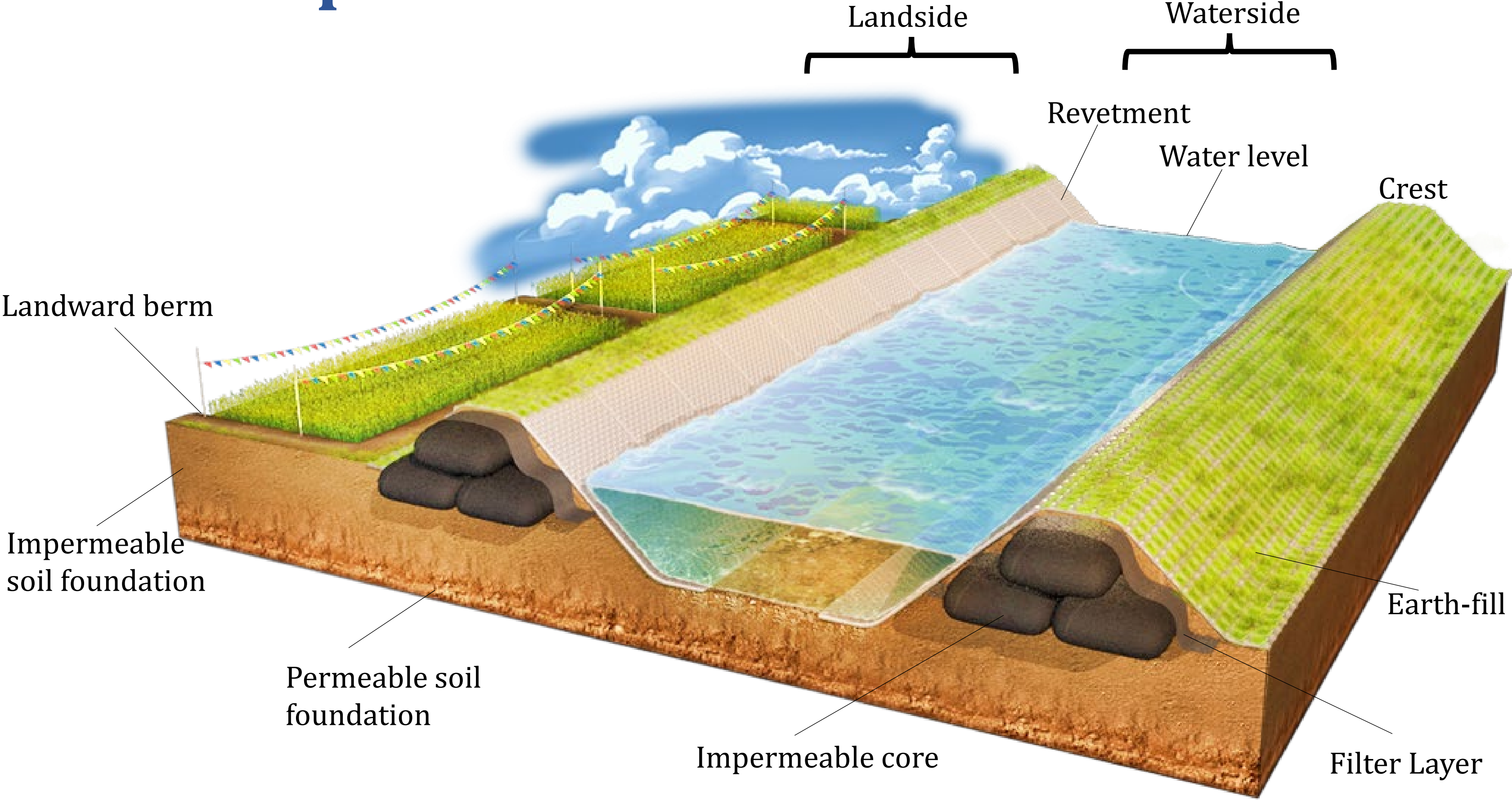
Source: Allsop et al. (2009)

Structural Approach : Levee / Embankment

- Levees are raised, predominantly earth, structures that are not reshaped under normal conditions by the action of waves and currents.
- Primary objective is to provide **protection** against fluvial and coastal flood events along coasts, rivers and artificial waterways.
- Unlike engineered structures, levees can be **irregular in the standard** and nature of their construction and **can deteriorate** markedly over time if they are not well maintained.



Main Components of Levees

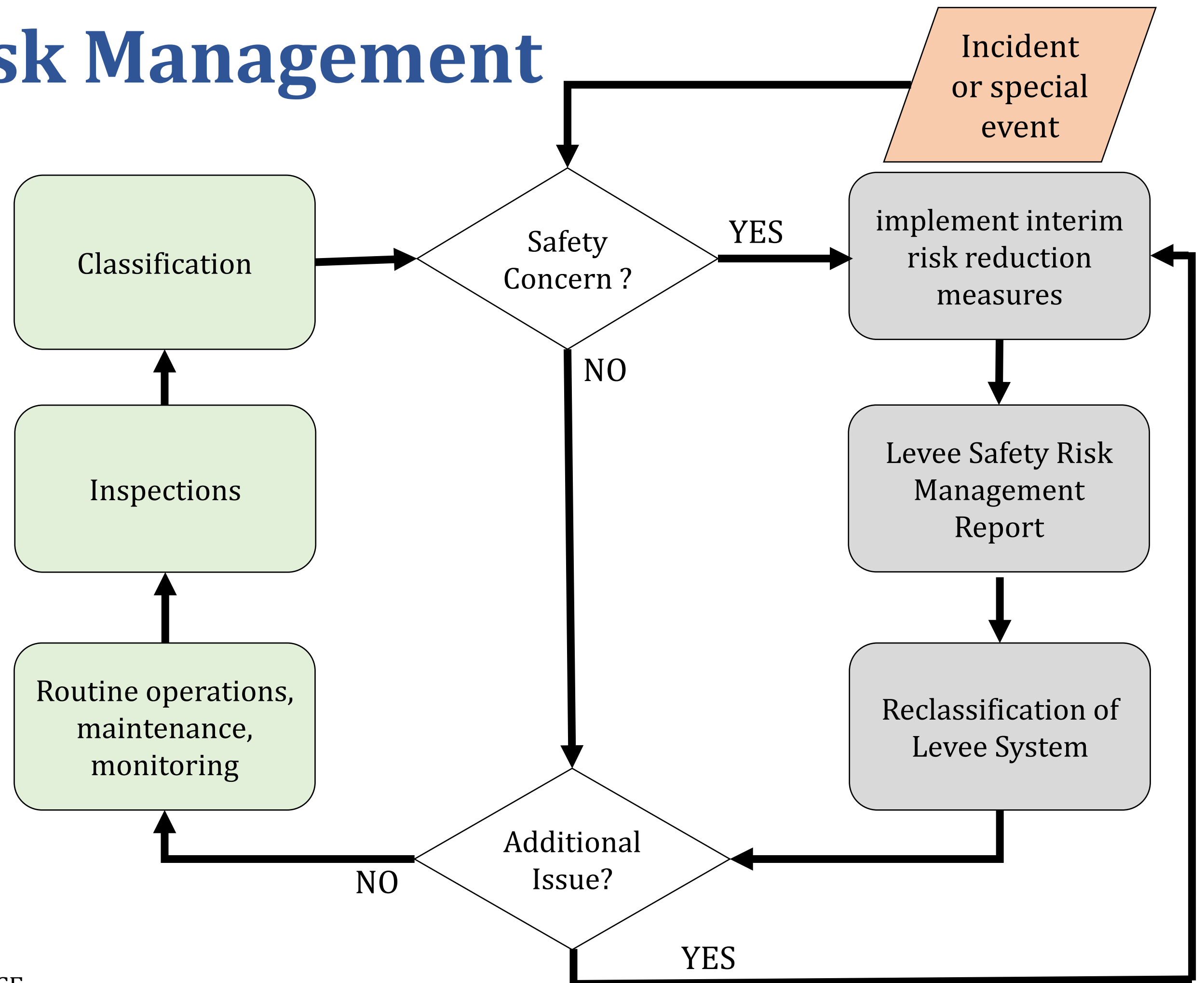


Source: adapted from <https://www.geoace.com/app/Riverbank-and-Channel-Protection/Levees-and-Dikes>

Levee Safety Risk Management

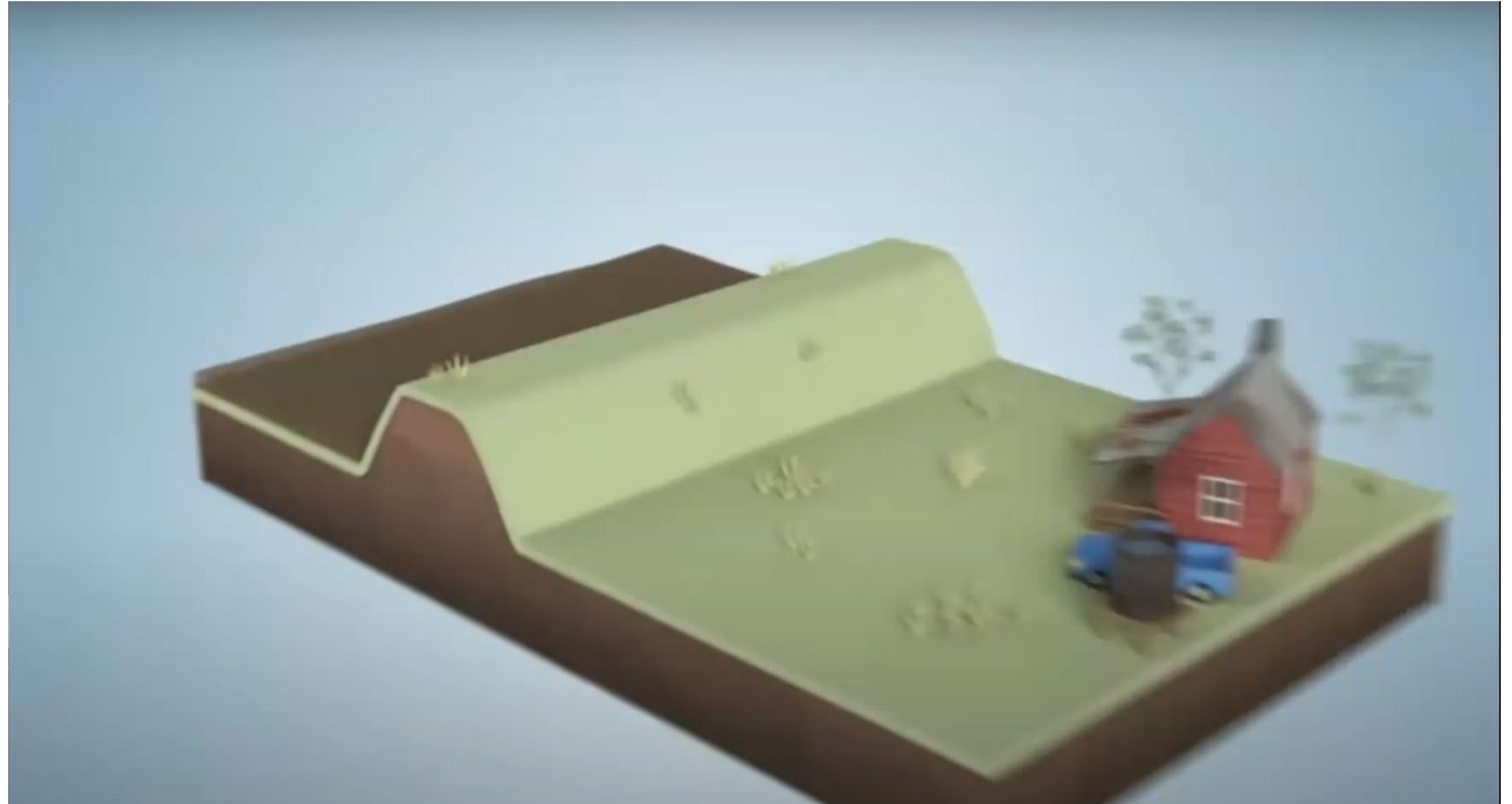
Key Point

*Typical routine activities are combination between **field inspections** and **screenings**. It must be remembered that although **flood risk may be reduced** by such an approach, **it can never be removed completely**.*



Disaster Potential of Embankment System

1. Overtopping
2. Cracking
3. Piping, Tree, Weather, and Animal/Human Activities

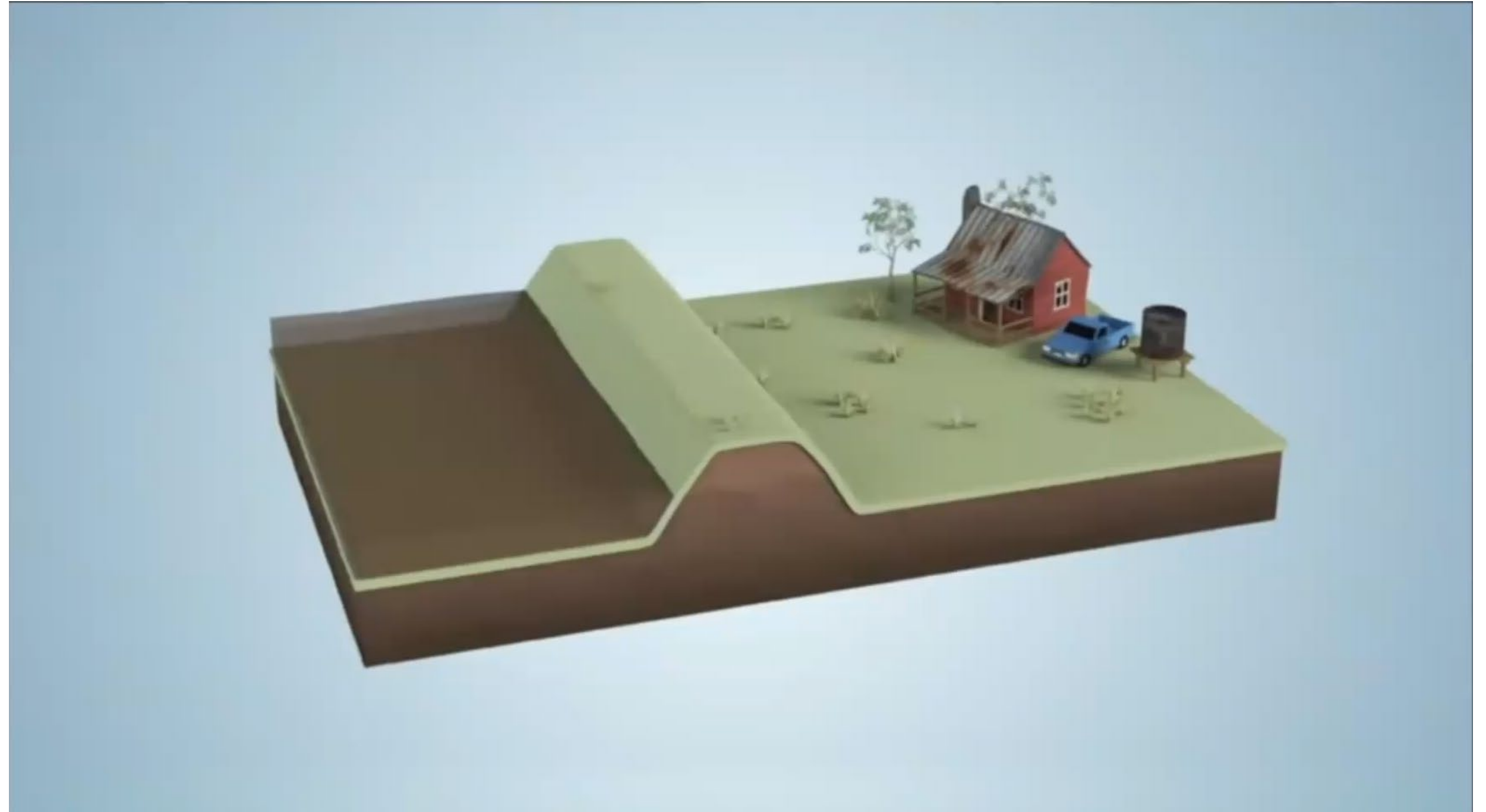


Disaster Potential of Embankment System

1. Overtopping

2. Cracking

3. Piping, Tree,
Weather, and
Animal/Human
Activities



Disaster Potential of Embankment System

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Disaster Potential of Embankment System

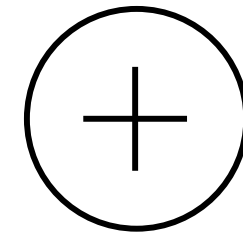
1. Overtopping
2. Cracking
- 3. Piping, Tree, Weather, and Animal/Human Activities**



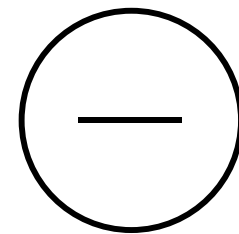
Main Challenge in the Embankment System

Key Point

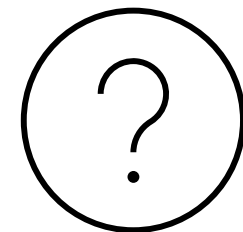
It is essential for mitigating flood embankment before extreme flood events and ensuring sustainable flood and coastal defence [1].



The rate of deterioration can be detected in **high accuracy** through long-term observation and inspection.



The traditional inspection methods by visual monitoring are **inefficient** and **inaccurate**.



The use of **algorithms** and techniques based on **remote sensing** can help local government to identify vulnerable levee sections and repair them rapidly with lower costs [2].

Goal and Objective

Key Point

*“The structure and components of the embankment should be in a **sustainable design**, consider the effect of **water loading**, and understanding **climate variability**.”*

Goal

This research aims to calculate the robust deterioration rate of soil embankment, particularly along the Thames and Humber Rivers, and predict future potential failure.

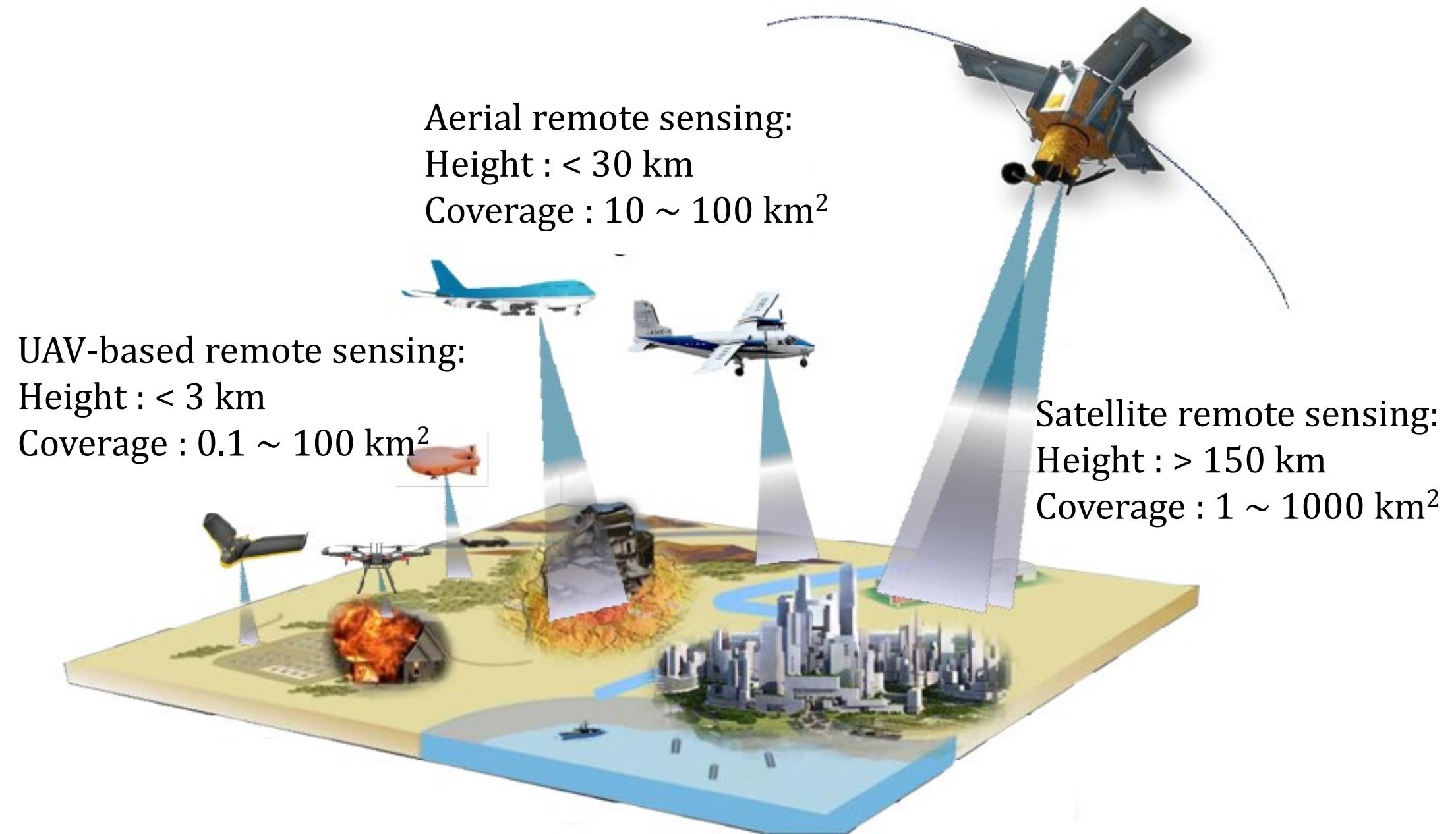
Objective

- To calculate deterioration rate of the embankment by identifying the sign of failure (subsidence, crack, or others)
- To project future potential of the embankment structure considering climate change approach

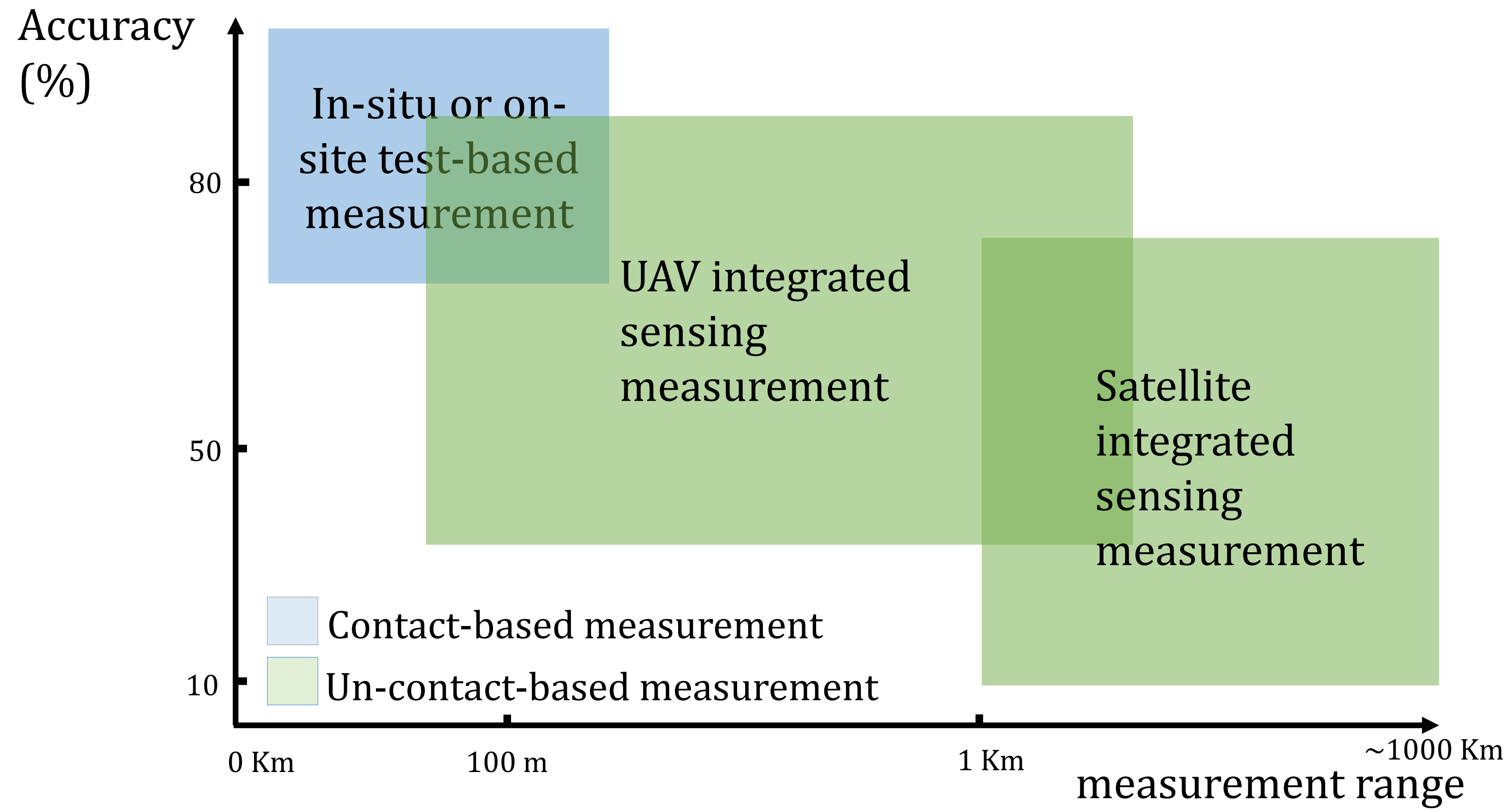
Unmanned Aerial Vehicle (UAV)

Advancements in sensor and UAV technologies have facilitated :

- more **intelligent monitoring** and inspection of sites prone to failure.
- provide both **in situ and satellite measurements** with sufficient accuracy.
- fewer **spatiotemporal constraints** and offer **superior resolution** with reduced data gaps compared to satellite-based measurements.



Comparison of UAVs Sensing Measurement



Source: Kim et al., 2023, Colomina et al., 2014 (UAV), Lillesand et al., 2015 (On-site Test), Congalton et al., 2009 (Satellite)

Previous Research

3D point cloud data derived from LiDAR is widely used in the sector of autonomous driving, satellite remote sensing, and spatial mapping

Key Point

LiDAR output will contain noise and non-target information due to the complex and changeable environment (weather, interference, other types of obstacles)

Note

Airborne/spaceborne LiDAR commonly uses scanning or array detection to obtain 3D point information. It is commonly resulting in low density and poor-quality point cloud data acquisition after scanning due to volume and weight data.

Solution

A variety of pre-processing processes after acquisition is important, such as point cloud segmentation, denoising, background point cloud removal, sparse/ missing point cloud completion.

Previous Research

Key Point

LiDAR Enable 3D landslide analysis by using a narrow laser beam to scan a certain object [1]

LiDAR can be used to monitor landslides, rockfalls, and debris flows by generating an accurate slope map [2], calculating instability signs of landslide [3], and understanding the landslide processes and reducing related losses [4]

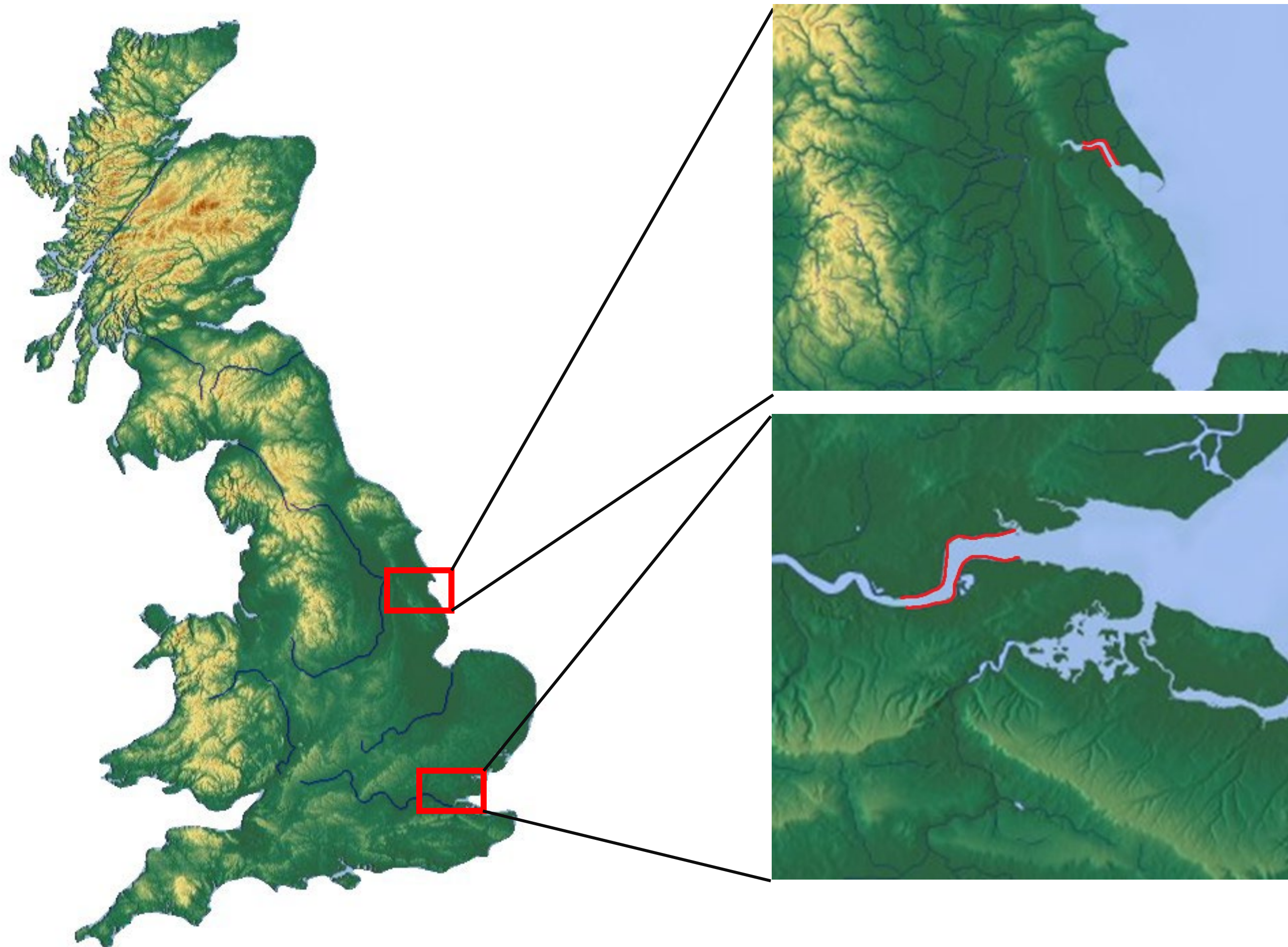
Note

Main factor of the quality of the application results of the LiDAR is **spatial resolution**, including sensor equipment performance, measurement distance, vegetation canopy density, and the **filtering algorithm** effect of the point cloud.

Solution

- Develop mathematical calculation as an input in computational methods that enable to identify potential failure on embankment system
- Develop a statistical approach to project future potential disaster under climate change consideration

Study Area



River Humber

The Humber is about 64 km long, the River is lined by the major ports of Kingston upon Hull, Grimsby, and Immingham.

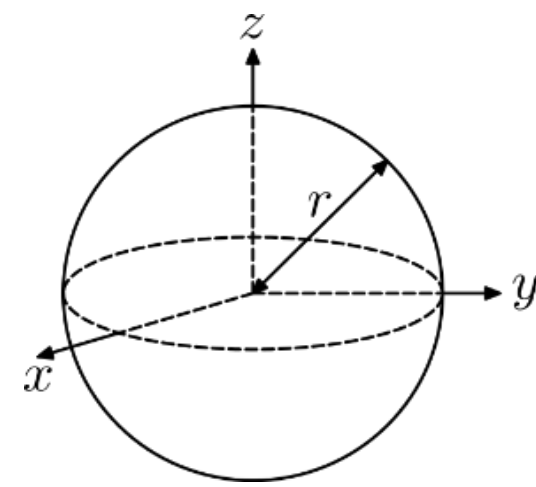
River Thames

The Thames is the largest river in England, with a total length of 354km, housing a fifth of the UK population, including London.

LiDAR

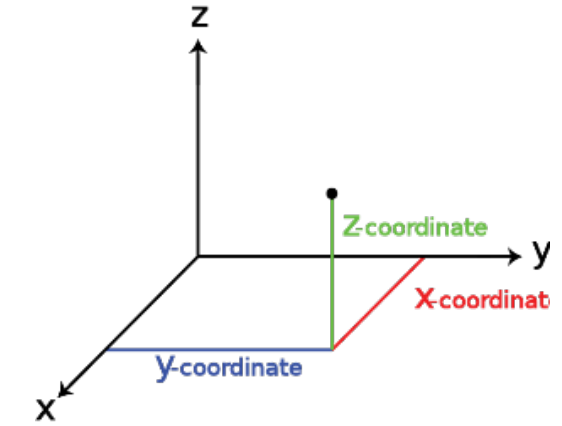
Key Point

LiDAR excels in capturing data with both **high spatial** and **spectral resolutions**, facilitating the generation of **precise classifications**, detection of surface alterations, environmental monitoring, and various other applications.



The IMU (inertial measurement unit) gives the precise orientation of the scanner

Drone equipped with LiDAR unit



The GPS gives the precise location of the scanner

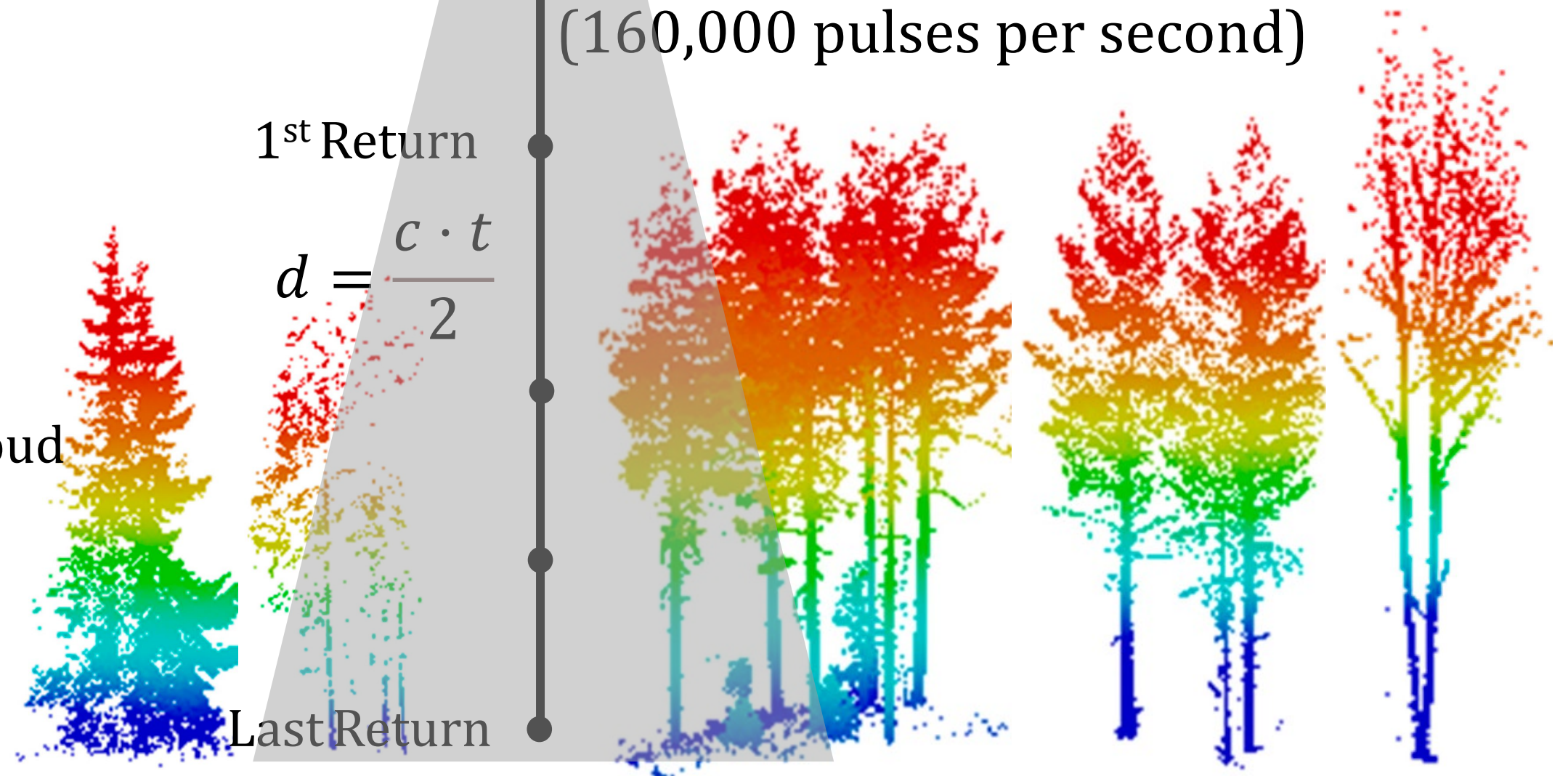
Laser Pulse
(160,000 pulses per second)

1st Return

$$d = \frac{c \cdot t}{2}$$

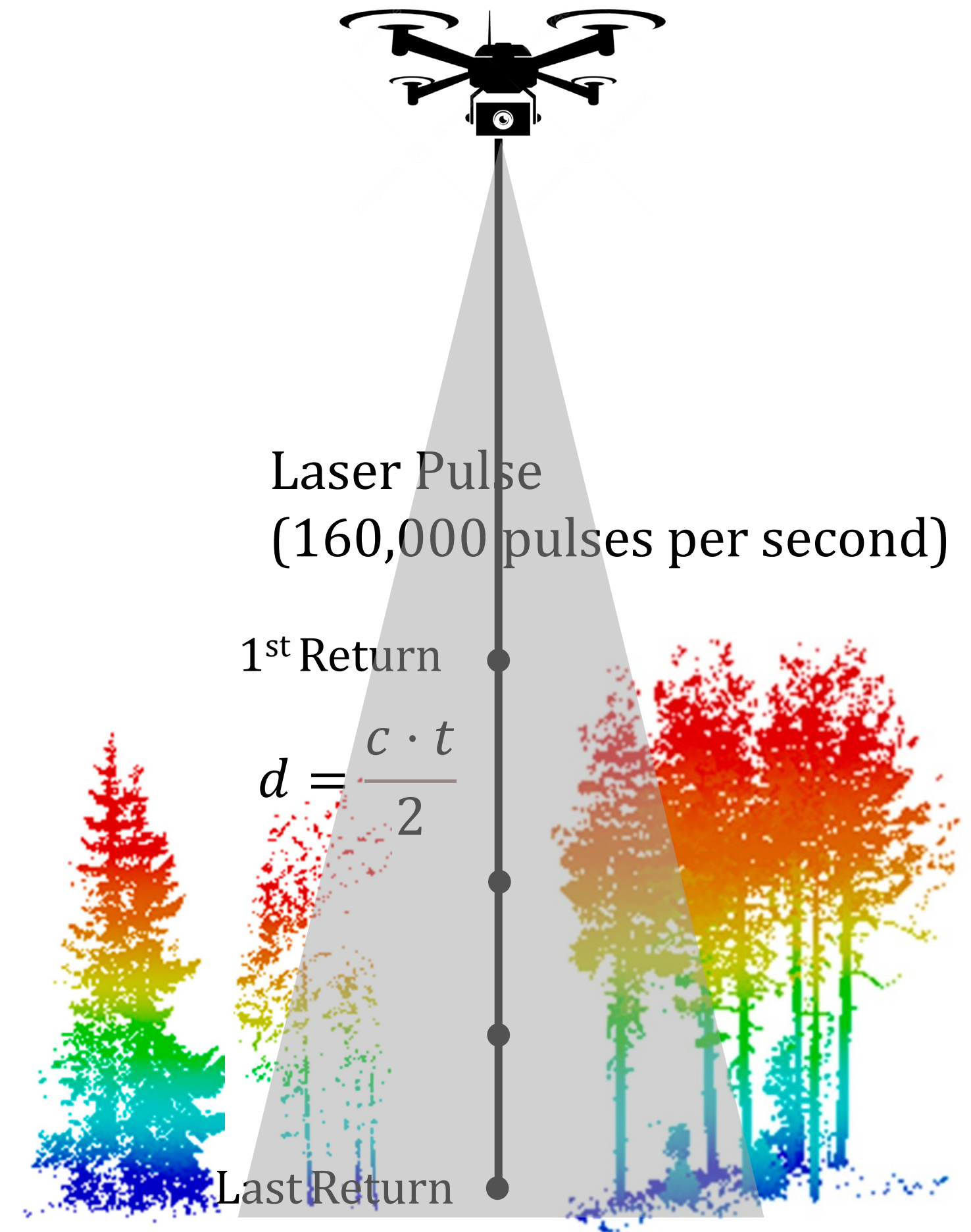
3D Point Cloud

Last Return



LiDAR

- One pulse may record 1-5 return pulses
- The returned pulses is classified into one or more discrete returns X, Y, Z intensity
- Optical frequency is between Green and near Infra-Red (wavelengths from 532 to 1064 nm)
- Spatial resolution is a function of the altitude and flight speed ranging from a few cm for altitudes of about 250 m to a few decimetres for higher than 1000m
- Can operate both day and night, some limitations may occur



LiDAR Limitation

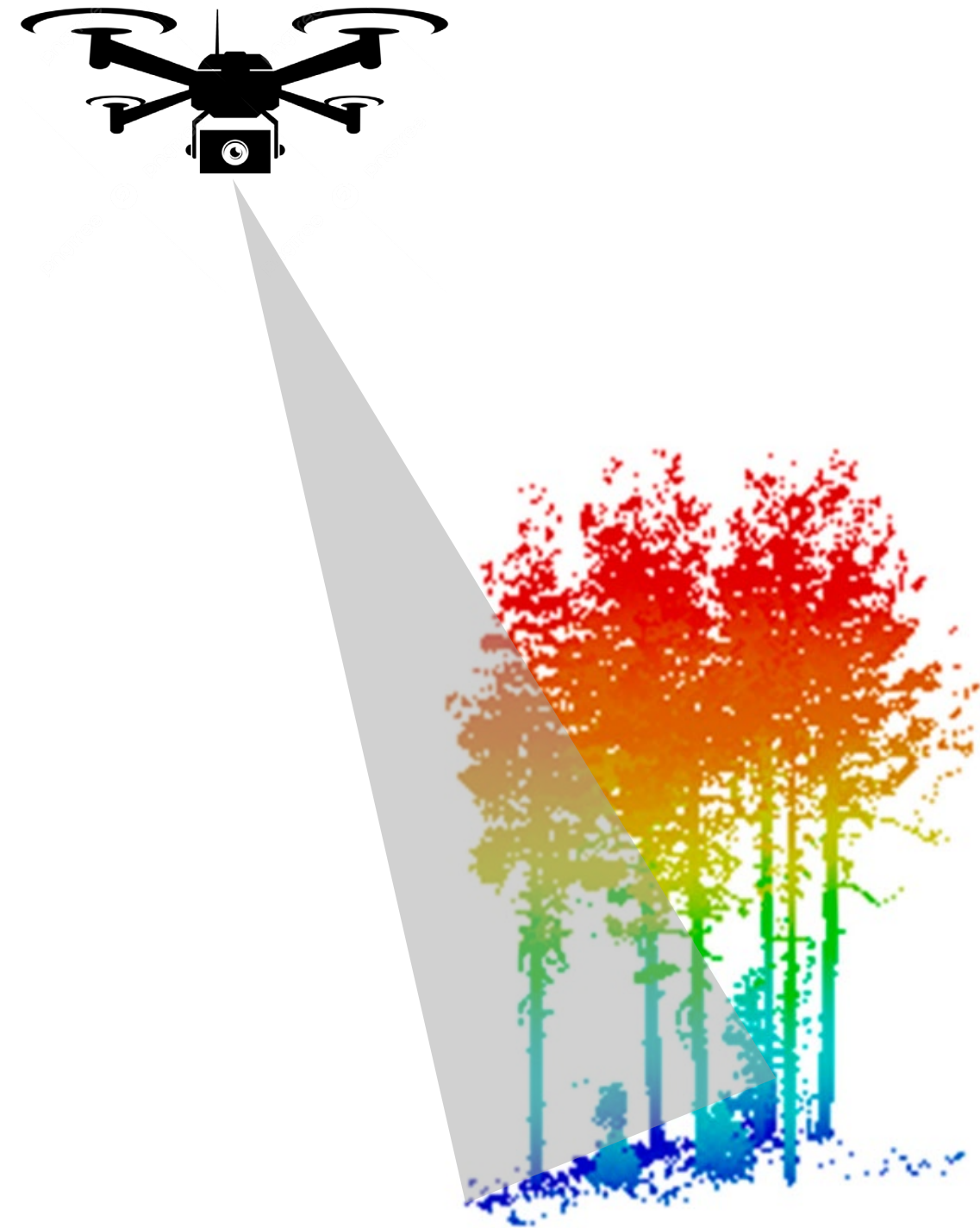
Obstruction




Weather



Angle



Data for Pre-Trained Method



Department for Environment
Food & Rural Affairs

Data Services Platform

[Create an account](#) [Login](#)

Home

APIs

App gallery

Surveys

Support

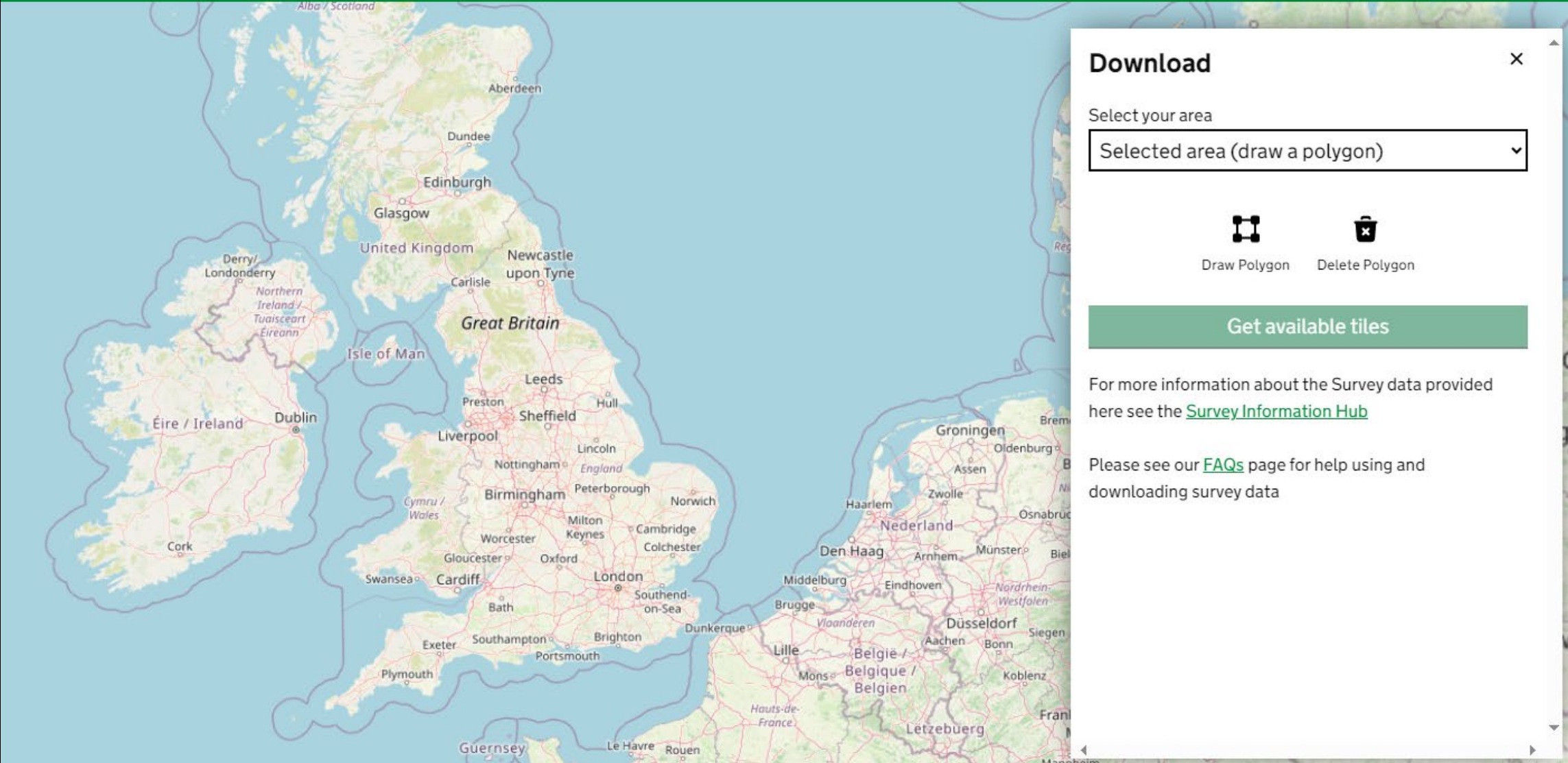
BETA

Contact the Data Services Platform Service team if you have feedback, questions or suggestions.

Defra Survey Data Download

Layers

Download



Download

Select your area

Selected area (draw a polygon)

Draw Polygon

Delete Polygon

Get available tiles

For more information about the Survey data provided here see the [Survey Information Hub](#)

Please see our [FAQs](#) page for help using and downloading survey data

Download

Select product

LIDAR Composite DTM

LIDAR Composite DTM

LIDAR Composite First Return DSM

LIDAR Composite Last Return DSM

LIDAR Point Cloud

LIDAR Tiles DSM

LIDAR Tiles DTM

National LIDAR Programme DSM

National LIDAR Programme DTM

National LIDAR Programme First Return DSM

National LIDAR Programme Intensity

National LIDAR Programme Point Cloud

National LIDAR Programme VOM

SurfZone DEM 2019

Vertical Aerial Photography Tiles RGBN

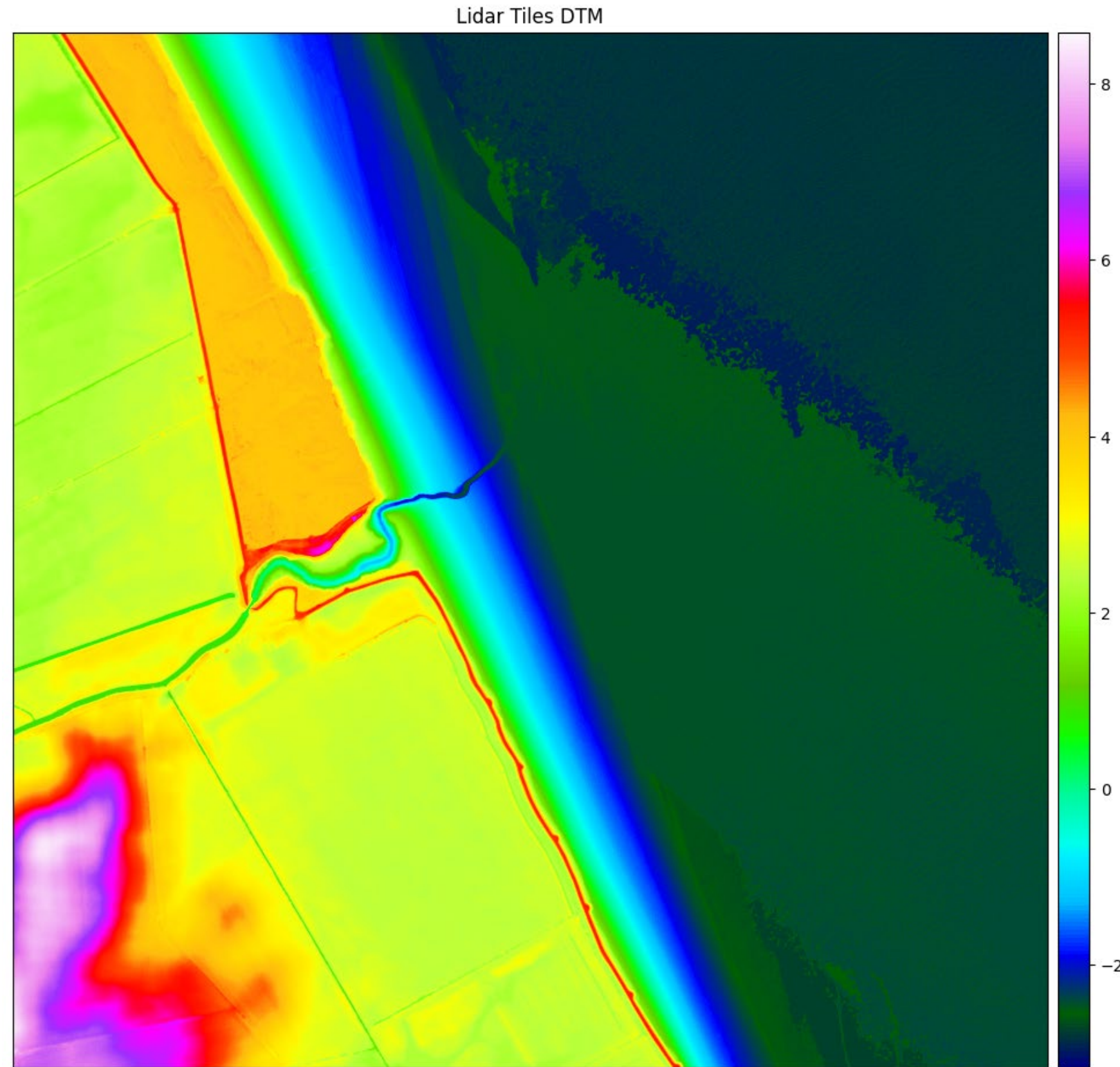
Link : [Defra Survey Data Download](#)

LiDAR Output

1. LiDAR DTM

2. LiDAR DSM

3. LiDAR Point Cloud



Digital Terrain Model

- Represent the elevation or relief of the Earth's surface, **exclude** human-made features and Trees
- Resolution and accuracy depending on the sensor, but commonly lies between 1 meter to 20 meter.

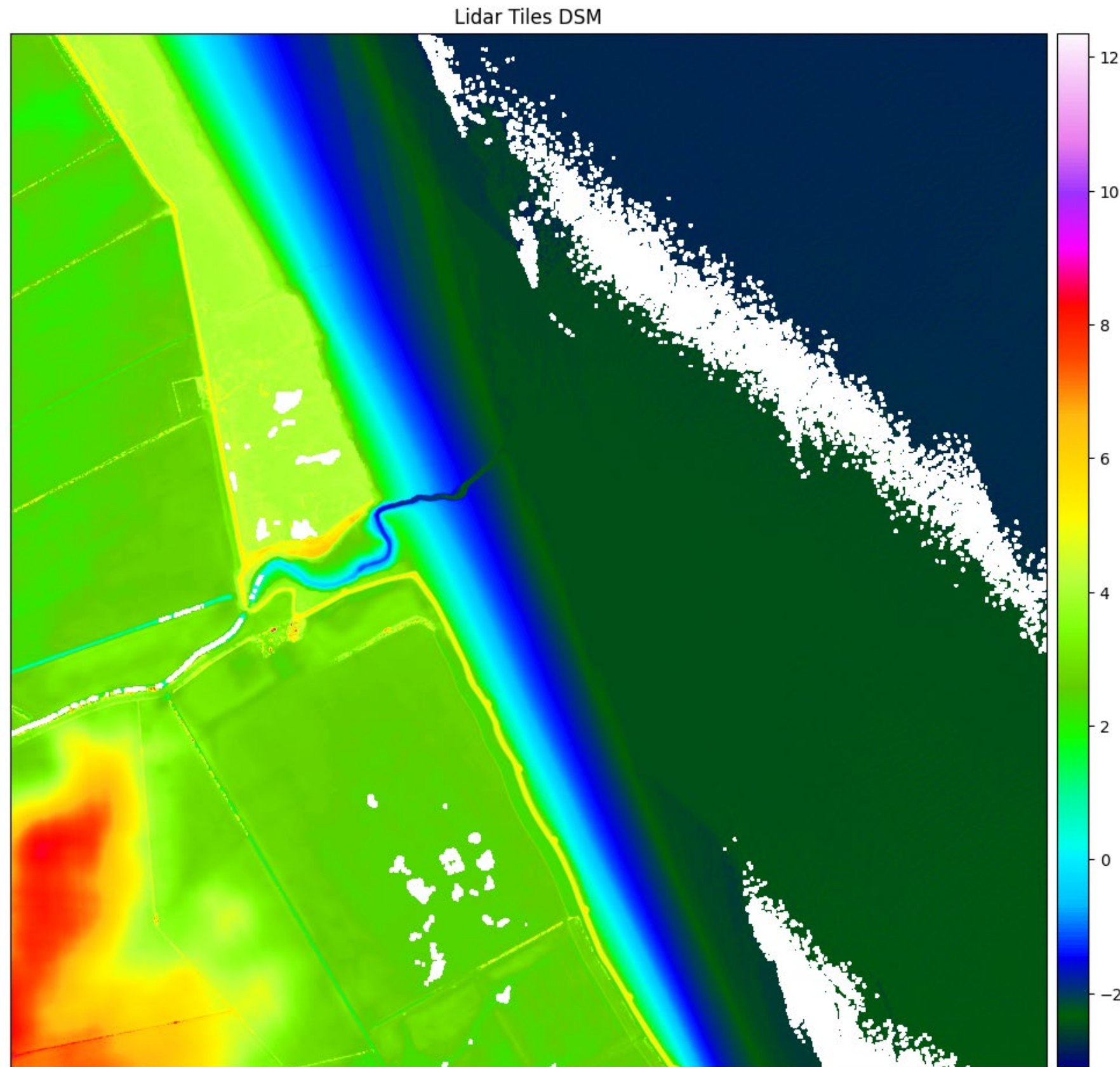
Source Data : DEFRA

Location : East Halton Skitter - Immingham

Tool : Google Colab - Python

LiDAR Output

1. LiDAR DTM
- 2. LiDAR DSM**
3. LiDAR Point Cloud



Digital Surface Model

- Represent the elevation or relief of the Earth's surface, **include** human-made features and surface features
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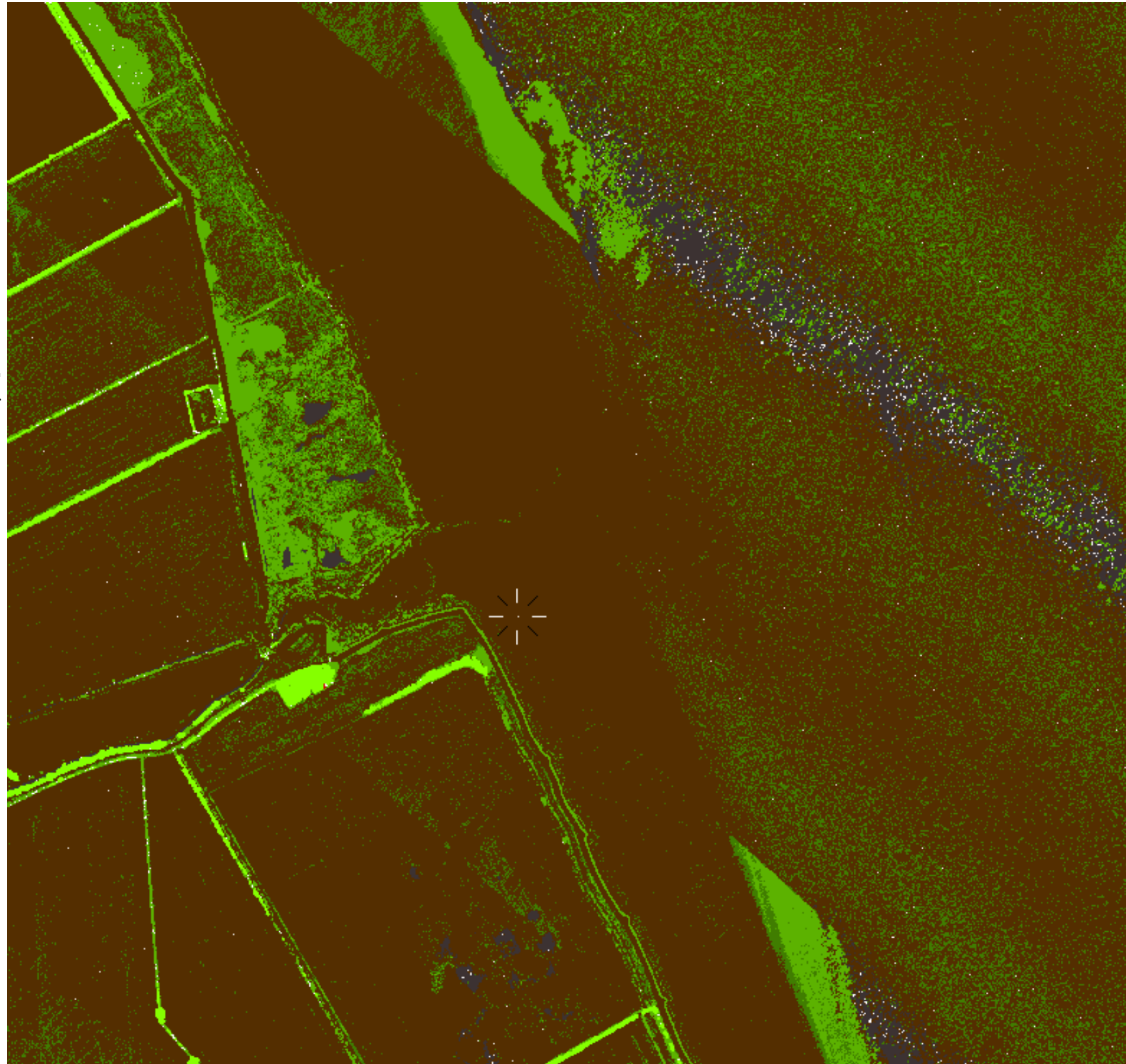
Source Data : DEFRA
Location : East Halton Skitter - Immingham
Tool : Google Colab - Python

LiDAR Output

1. LiDAR DTM

2. LiDAR DSM

3. LiDAR Point Cloud



Point Cloud

- Individual points in three-dimensional space
- Accuracy and precision depending on some factors (equipment, sensors, processing algorithm used)
- Can be generated from airborne or terrestrial LiDAR, laser scanning, or structured light scanning)

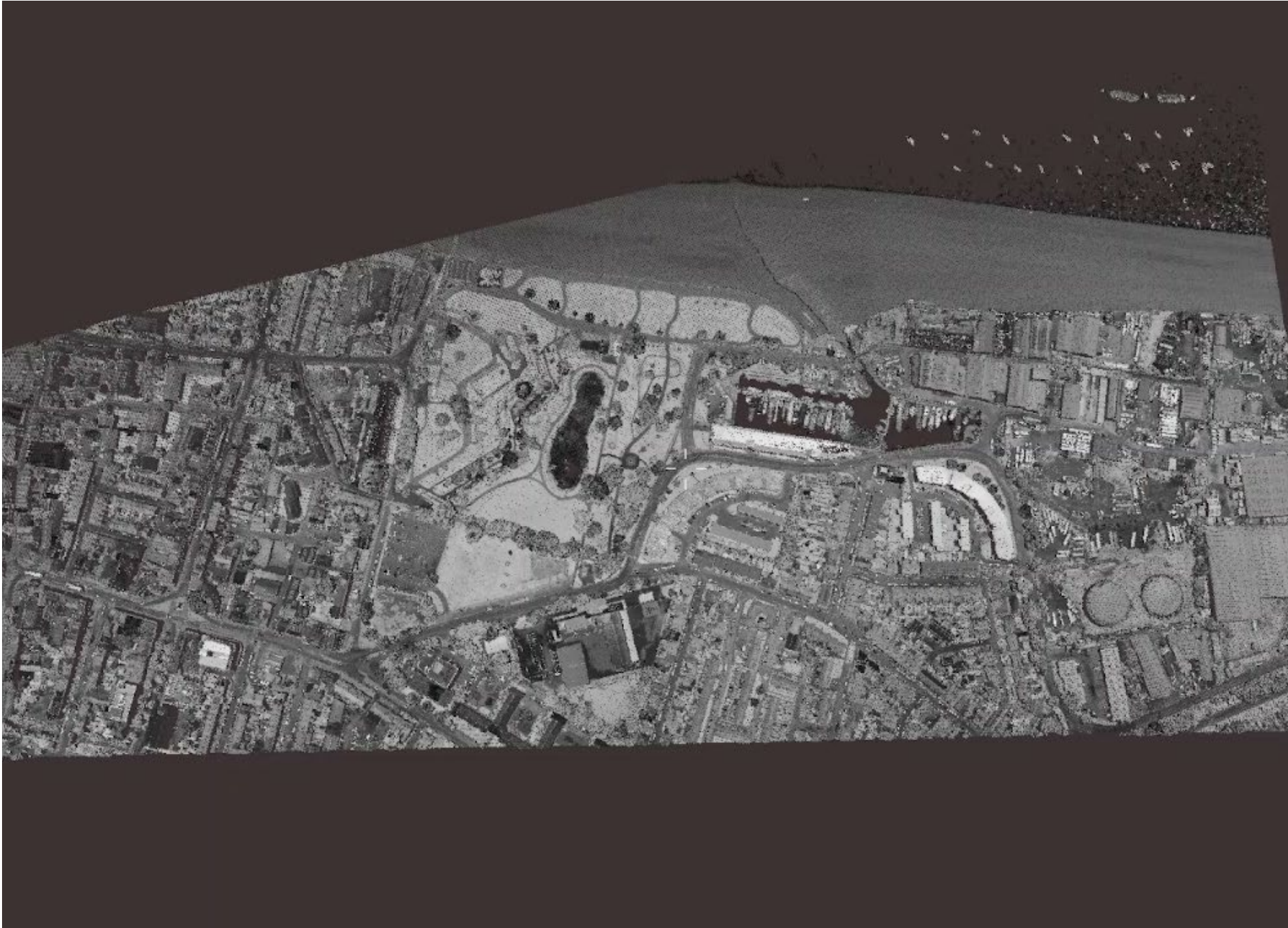
Source Data : DEFRA

Location : East Halton Skitter - Immingham

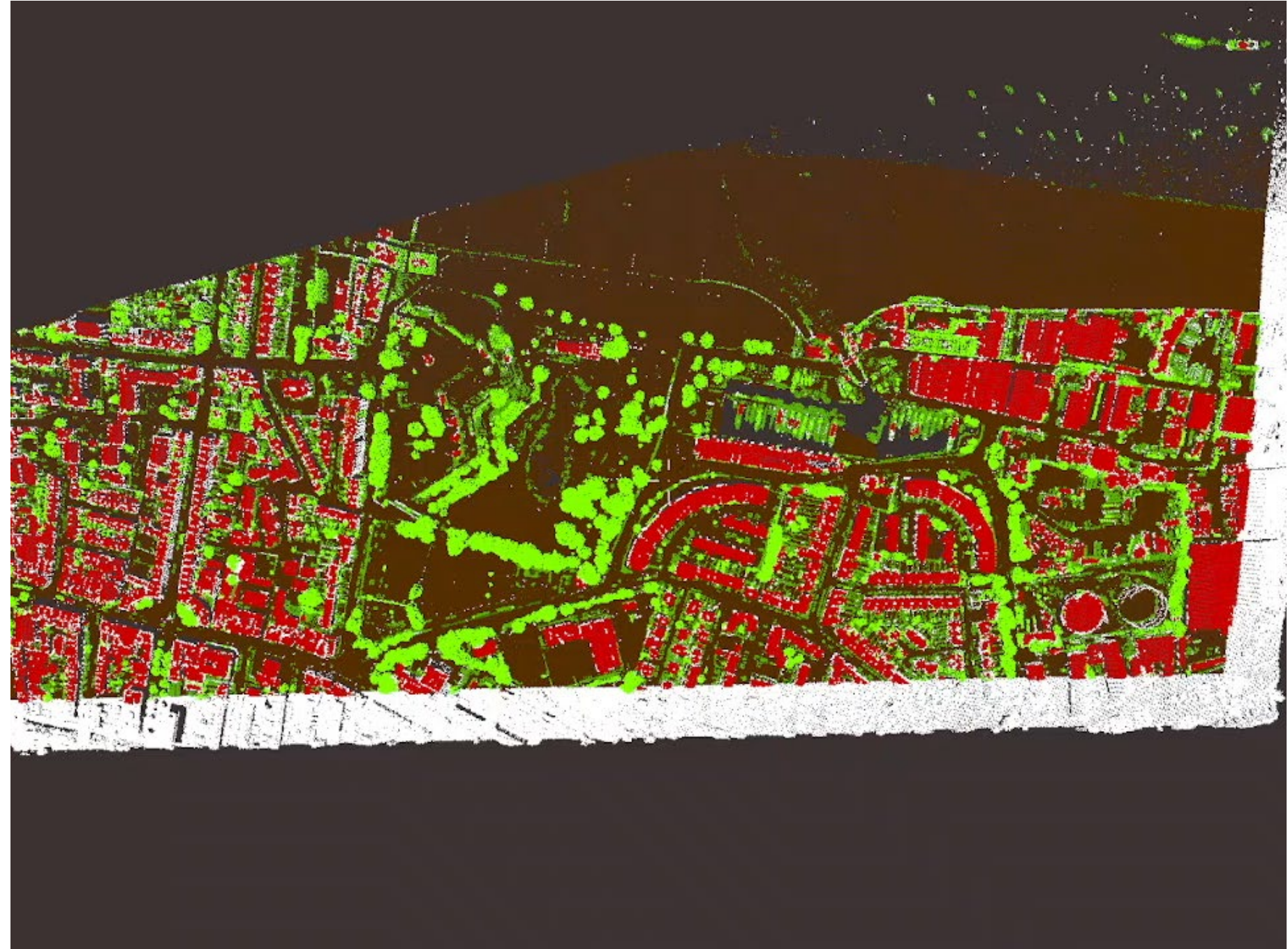
Tool : Displaz

LiDAR 3D Visualization

LiDAR Intensity Data



LiDAR Classification

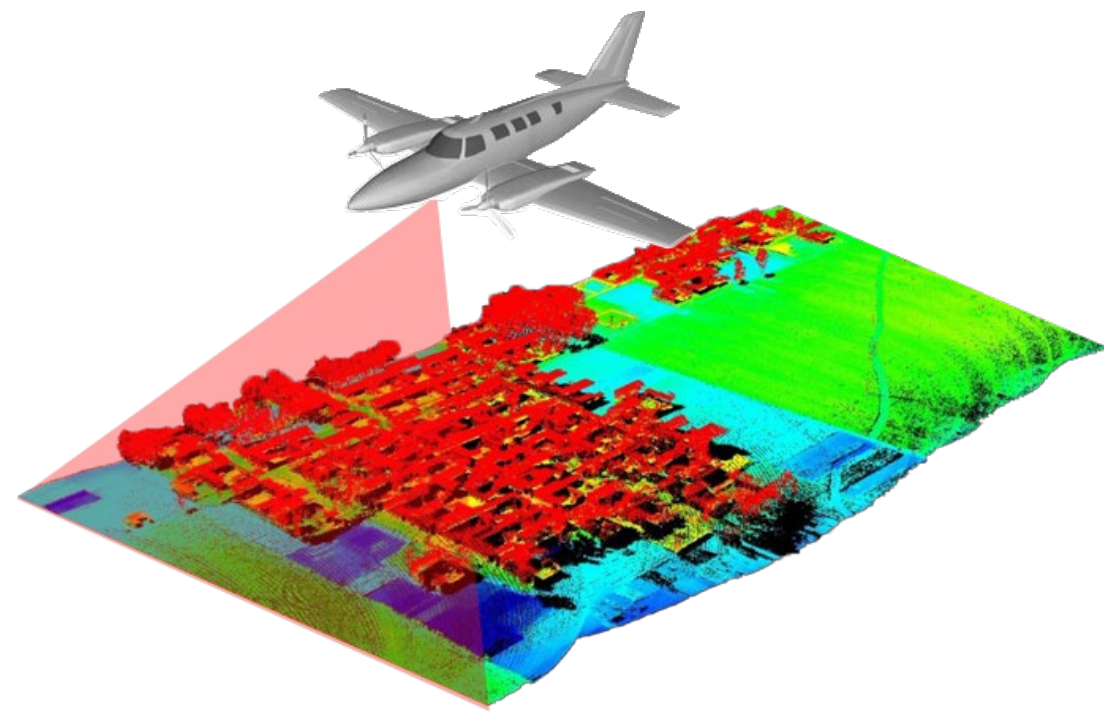


Source Data : DEFRA
Location : Residential Area in East Tilburn
Tool : Dislaz, CloudCompare

Milestone of Methodology

Lidar Data Acquisition and Pre-processing

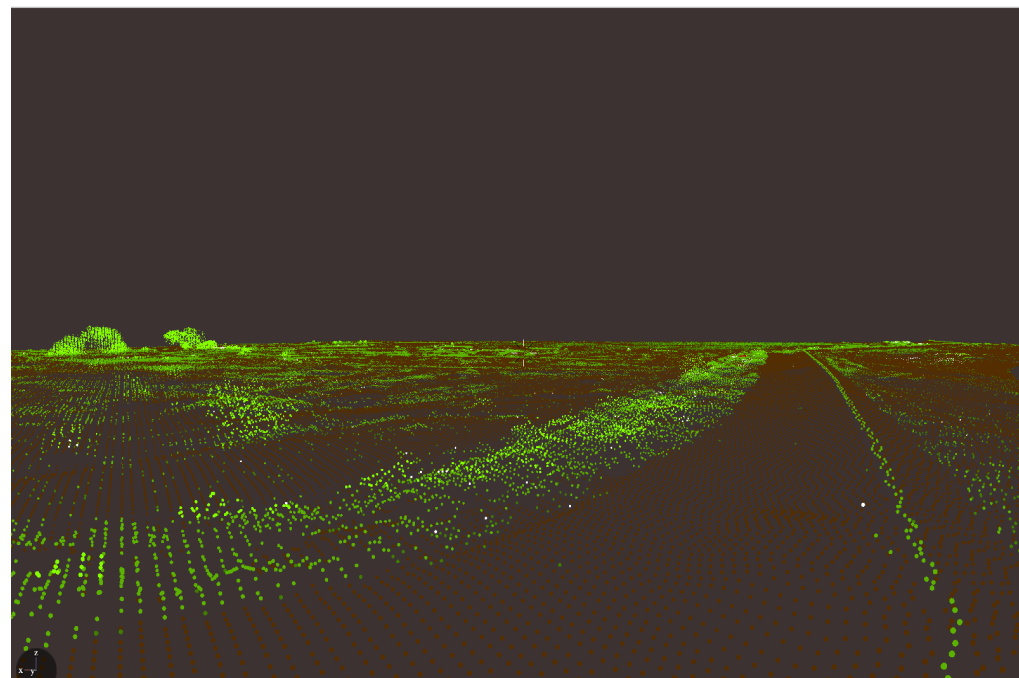
Gathering the data sources and comprehend the data typology.



Source: [LIDAR - lidar.co.id](http://lidar.co.id)

Segmentation and deterioration analysis

To isolate specific research areas and identify key features from the segmented data relevant to assessing the health and stability of flood embankments.



Source : LiDAR Data DEFRA
Tool : Displaz
Location : Lower Hope Point - Rochester

Predicting and Analysing Future Potential

To project the embankment deterioration rate by applying real-world scenarios to inform decision-making processes in civil engineering.

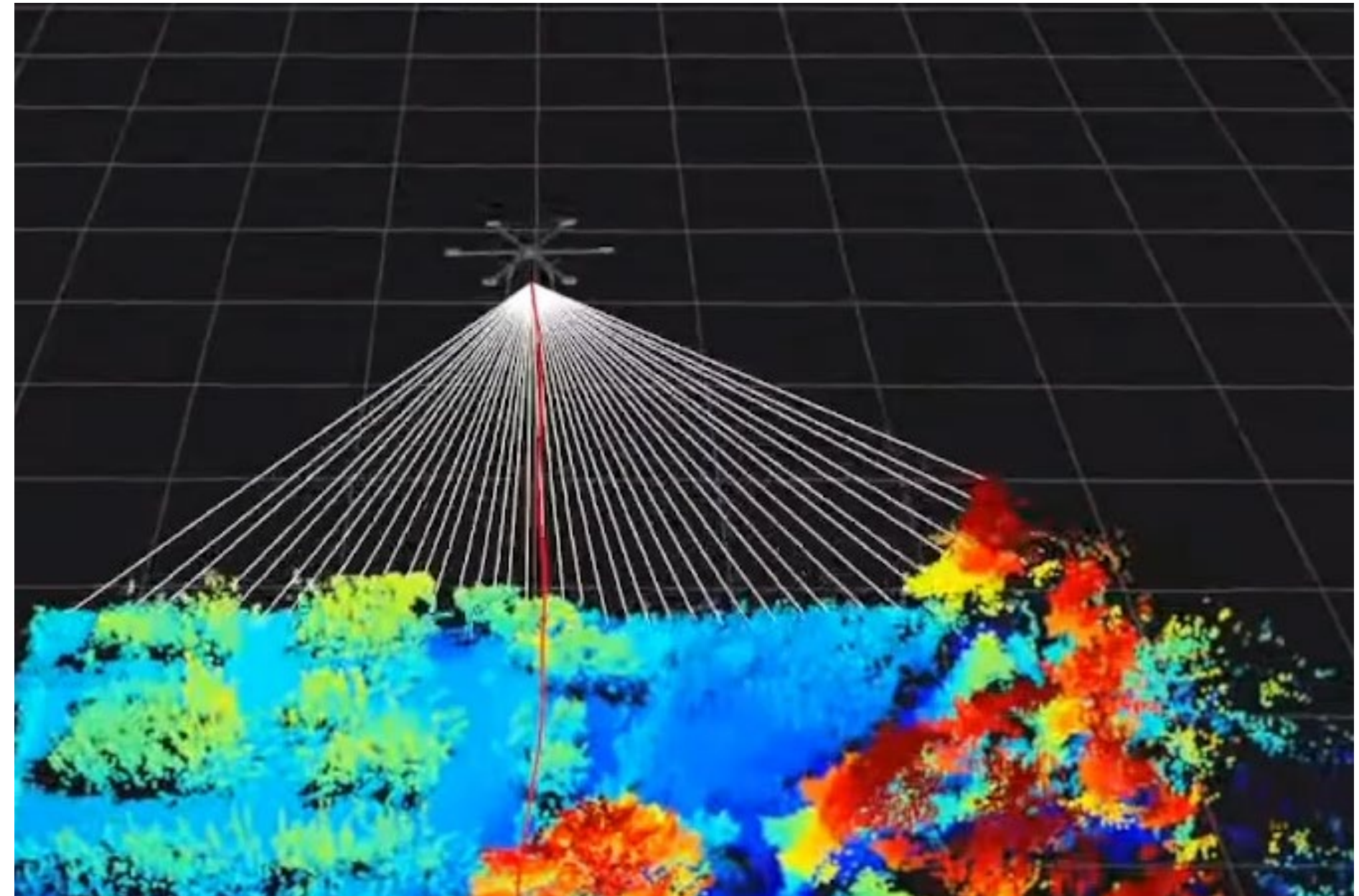


Source : Google Street View
Location : Lower Hope Point - Rochester

Lidar Data Acquisition and Pre-processing

Steps:

1. Project Planning and Preparation
2. Sensor selection and configuration
3. Platform deployment (aircraft / ground vehicle / tripod-mounted, or else)
4. Data acquisition
5. Quality control and assurance
6. Data processing



Segmentation and Deterioration Analysis

Key Point

A segment-based approach offers a systematic and effective methodology for analysing scalar field data representing surface features to detect embankment cracks

NOTE

Divided into segments or regions of interest based on geometric properties, such as curvature, normal vectors, or point density.

Application

Identifying and analysing specific object recognition, scene understanding, and semantic segmentation.

Advantage

Localised analysis, flexible, and interpretable

Drawback

Segmentation sensitivity, over-segmentation, and computational complexity

Segmentation



Input point cloud

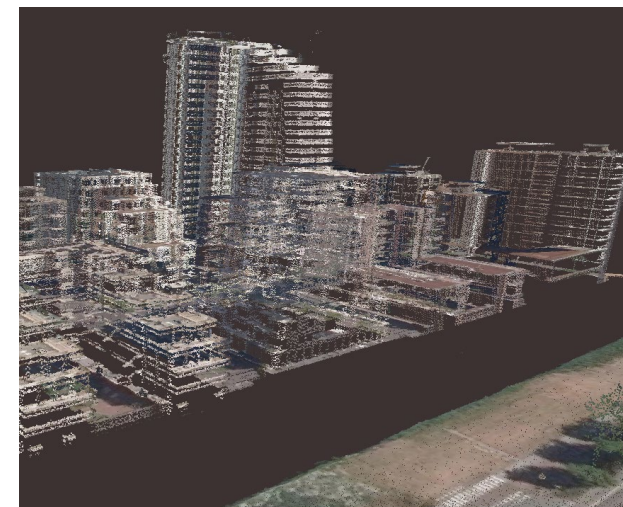
Select View



Surface
Rooftop
Trees
Embankment
River
Etc.

Image
Segmentation

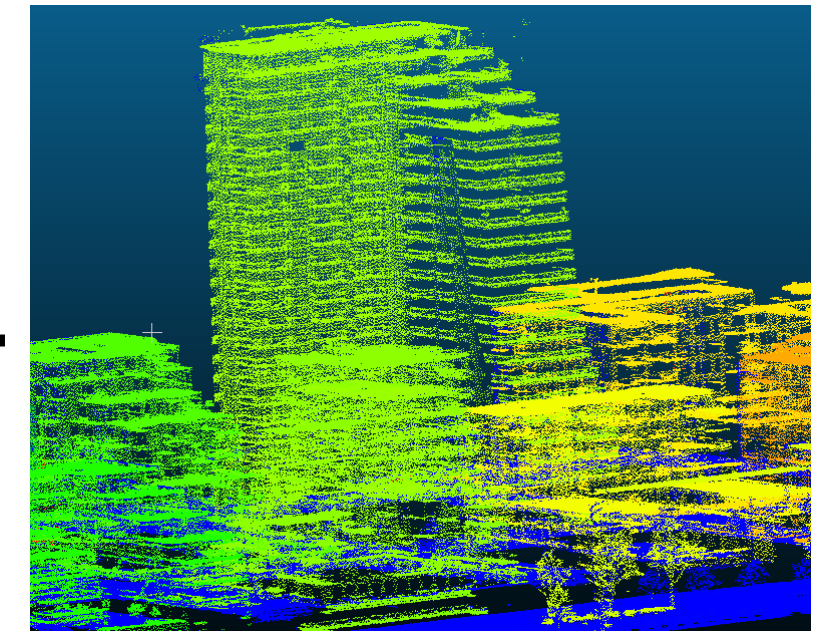
Pixel-Level label



render

Re-project

Point-Level label



Point
Segmentation

Output

Climate Projection

Challenge

No one single dataset in the public domain is properly detailed to describe for primary research. Water loss from subgrade soil subjected to dry weather is possible to cause soil cracking. On the contrary, intruded water will cause failures because of the wetting swelling or collapsing of the subgrade soils.

Key Point

3D point cloud data provided by a terrestrial laser scanner could play an interesting role for flood mapping.

Solution

Currently, the risk of floods is mapped on a global scale using technology like satellite imagery and remote sensing. LiDAR derived flood inundation model can be used to simulate flood hazard estimation using probability analysis and flood scenario.

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