

SBW AIDD: Developing an Adaptive and Intelligent Defoliation Detector

Adaptive deep and transfer learning to predict tree level SBW annual and cumulative defoliation at various platforms

SBW EIS Phase II Remote Sensing Project

TR 2022 No. 28

Request for the full report must be submitted to the author via: library@fpinnovations.ca

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Project number:301013892

ACKNOWLEDGEMENTS

This project was financially supported by Natural Resources Canada and industrial membership support to FPInnovations.

Brainstorming discussions with Denis Cormier (VP, FPI) and inputs to improve our understanding had been invaluable to the project.

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List of abbreviations used

AnnBr - current / annual defoliation estimated through branch / shoot assessment

AnnOc - current / annual defoliation estimated through ocular / binocular assessment

CB – Campbellton sites, NB

CHM – Canopy height model

CNN - Convolutional Neural Network

CS - Causapscal, QC

CumBr-cumulative defoliation estimated through branch / shoot assessment

CumOc – cumulative defoliation estimated through ocular / binocular assessment

CurBr- current / annual defoliation estimated through branch / shoot assessment

CurOc – current / annual defoliation estimated through ocular / binocular assessment

DEM - Digital elevation model

DSM – Digital surface model

GPS - Global Positioning System

Med CumBr- cumulative defoliation summarized as the median of the sampled shoots

Med CurBr – annual defoliation summarized as the median of the sampled shoots

MSS –Multi spectral sensor / camera acquiring data in red, green, blue, red edge and near infra red regions

NB - New Brunswick

NoDef - No defoliation (healthy, unaffected by SBW)

QC - Quebec

Res - image resolution

RGB – sensor / camera acquiring data in red, green and blue visible regions of the light spectrum

RTK – Real Time Kinematics

SD – Saint-Donat, QC

SMOTE – Synthetic Minority OverSampling Technique

SWIR - Short wave infra red

UAV - Unmanned Aerial Vehicle or commonly known as drone

VI – Vegetation index

BRDF – Bidirectional reflectance distribution function

TIR – Thermal Infrared

GCP – Ground control point

NDRG – Normalised difference of red and green

TGI – Triangular greenness index

VARI – Visible atmospherically resistant index

VIBR – Vegetation Index with ratio of blue to red

GNDVI – Green normalized vegetation index

NDVI – Normalized vegetation index

CMS -

GRVI – Green red vegetation index

LCI – Leaf chlorophyll index

NDRE – Normalized difference in rededge

VIGR - Vegetation Index with ratio of green to red



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