

# **Supporting Information for:** Applying a deep learning pipeline to classify low-quality historical RGB imagery into land cover classes

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## **1 Additional figures**

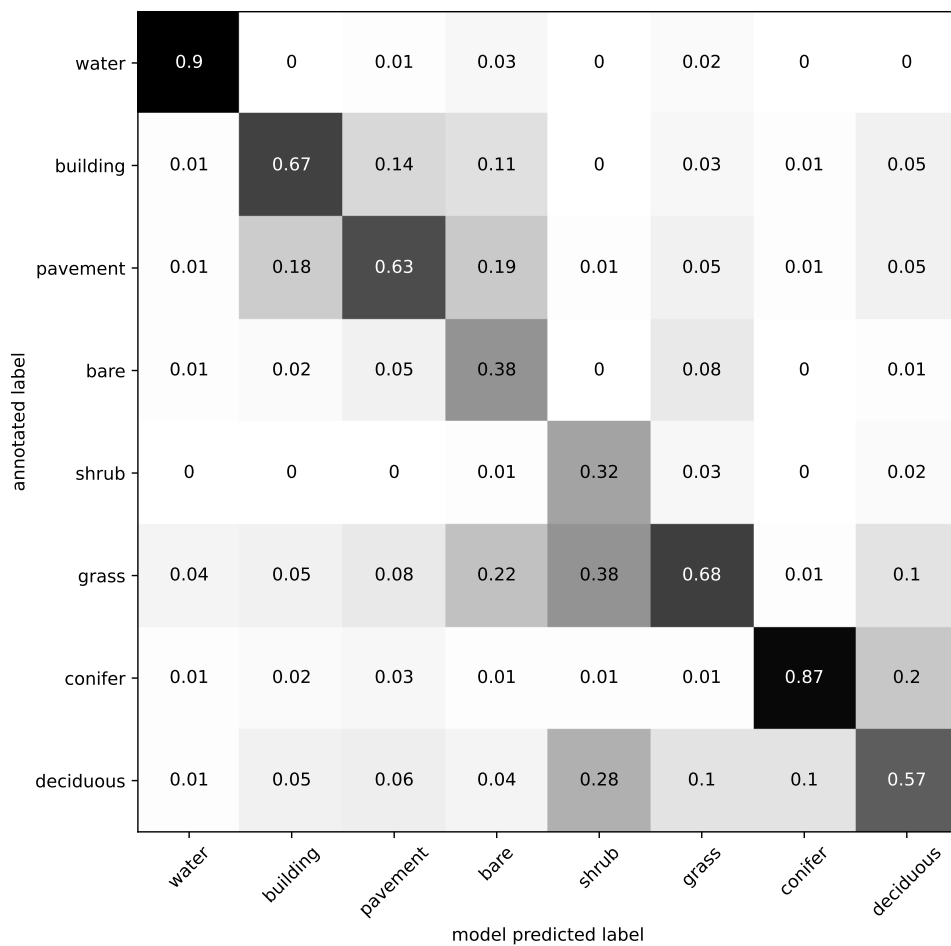


Figure S1: Modern confusion matrix, normalized by column.

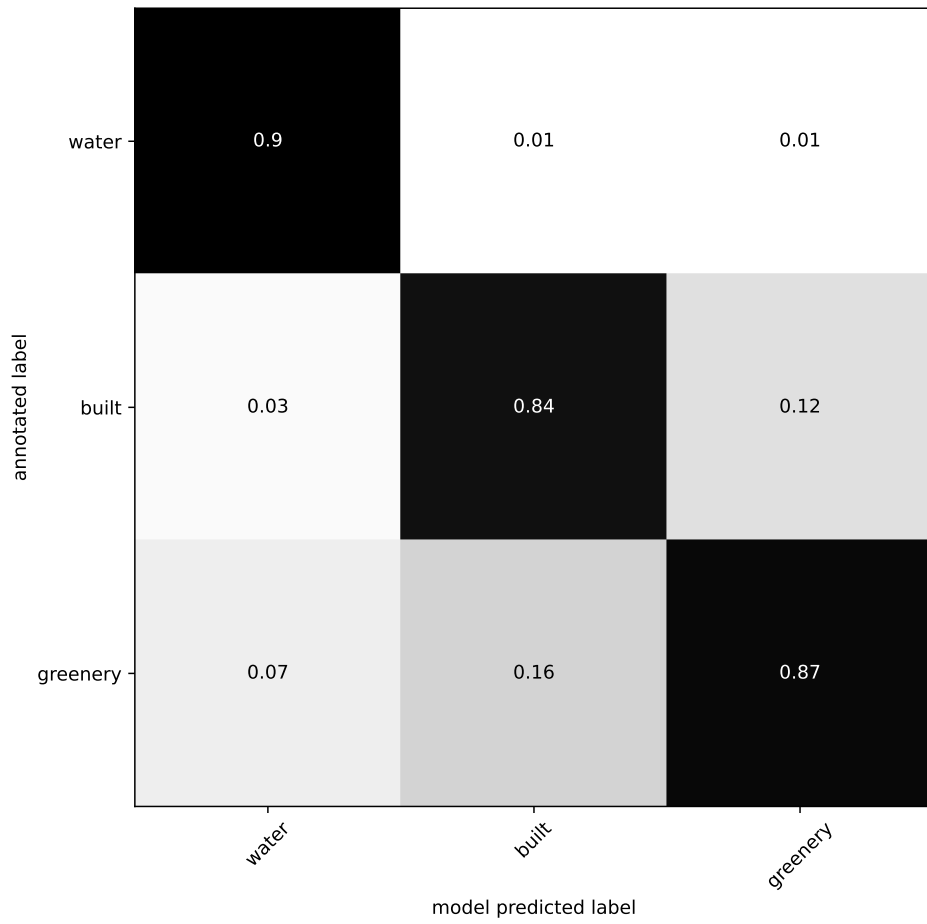


Figure S2: Modern confusion matrix after binning into water, built, and greenery categories, normalized by column.

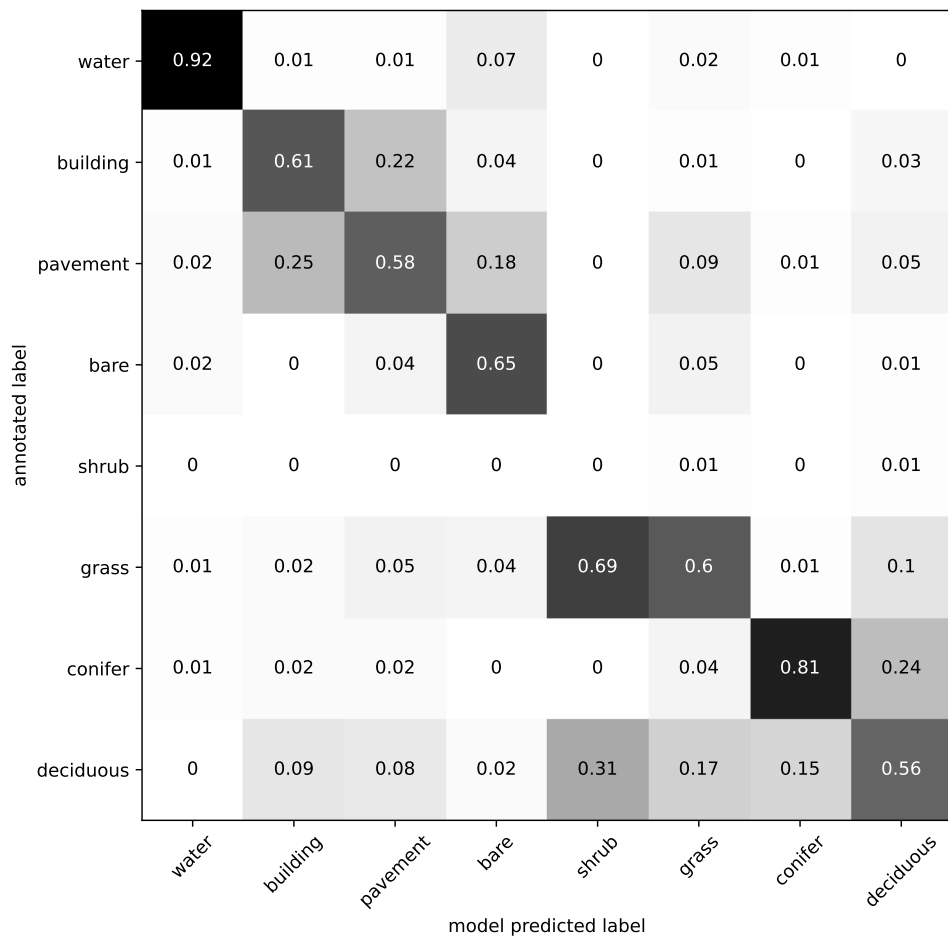


Figure S3: Historical confusion matrix after training on hand-annotated historical imagery augmented by rotation, normalized by column.

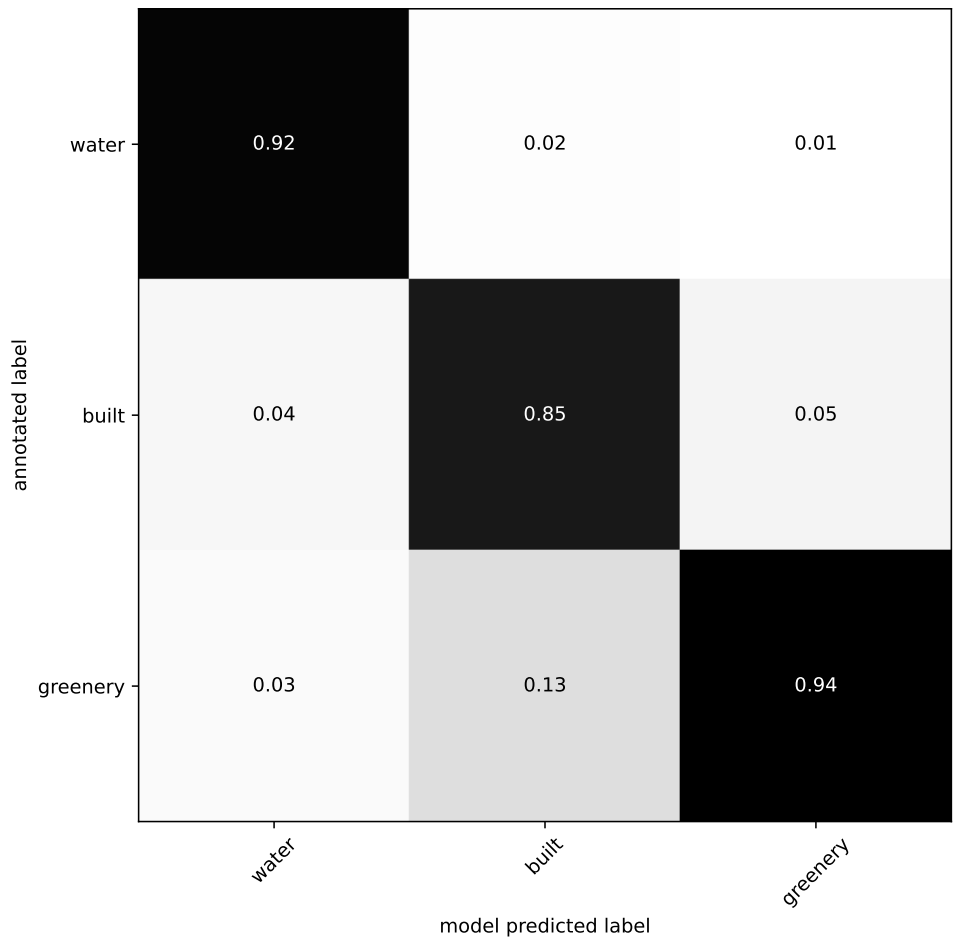


Figure S4: Historical confusion matrix after binning into water, built, and greenery categories and after training on hand-annotated historical imagery augmented by rotation, normalized by column.

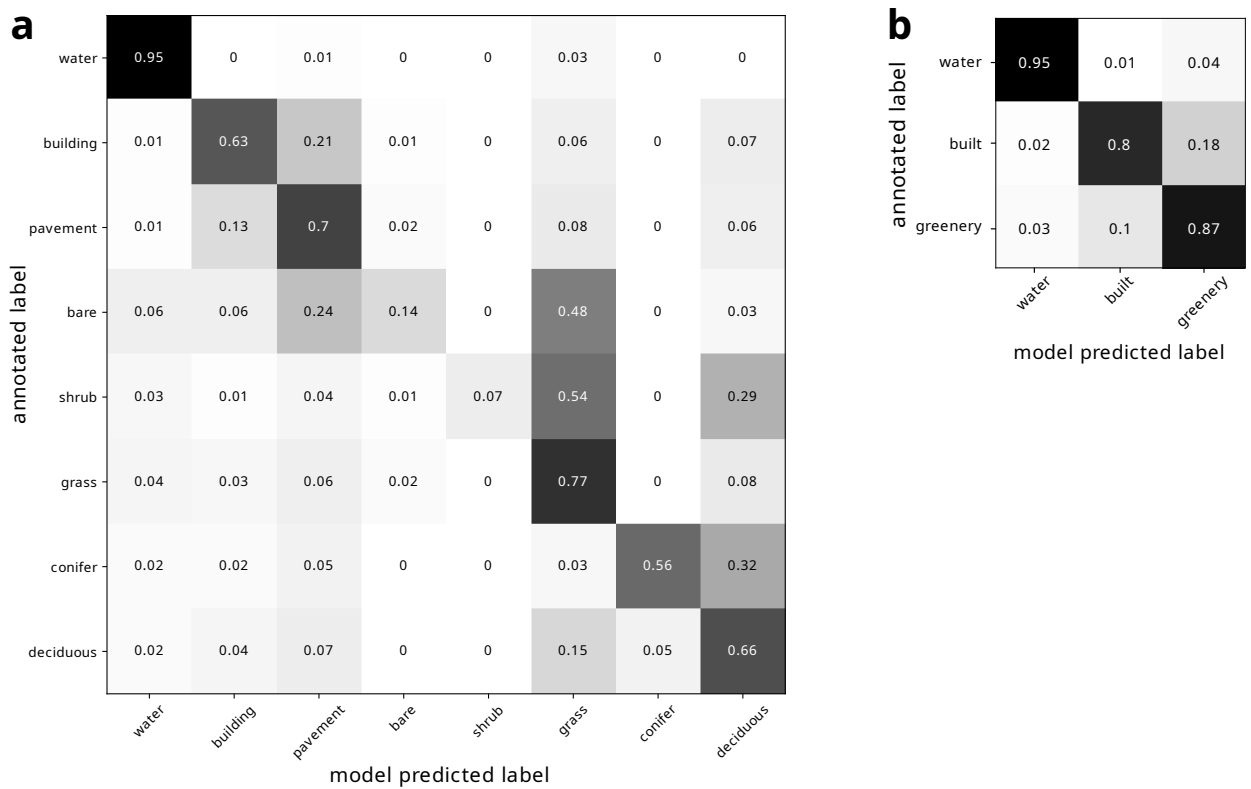


Figure S5: Confusion matrix of modern land cover model predictions, normalized by row of (a) all land cover types and (b) binned land cover categories

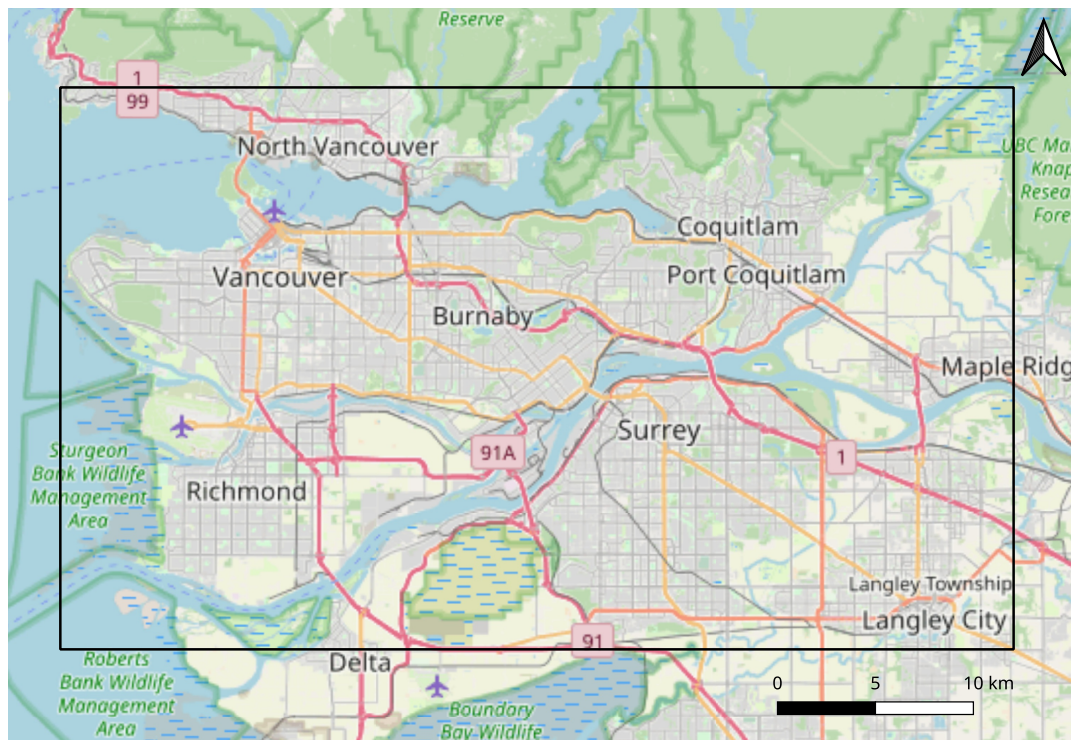
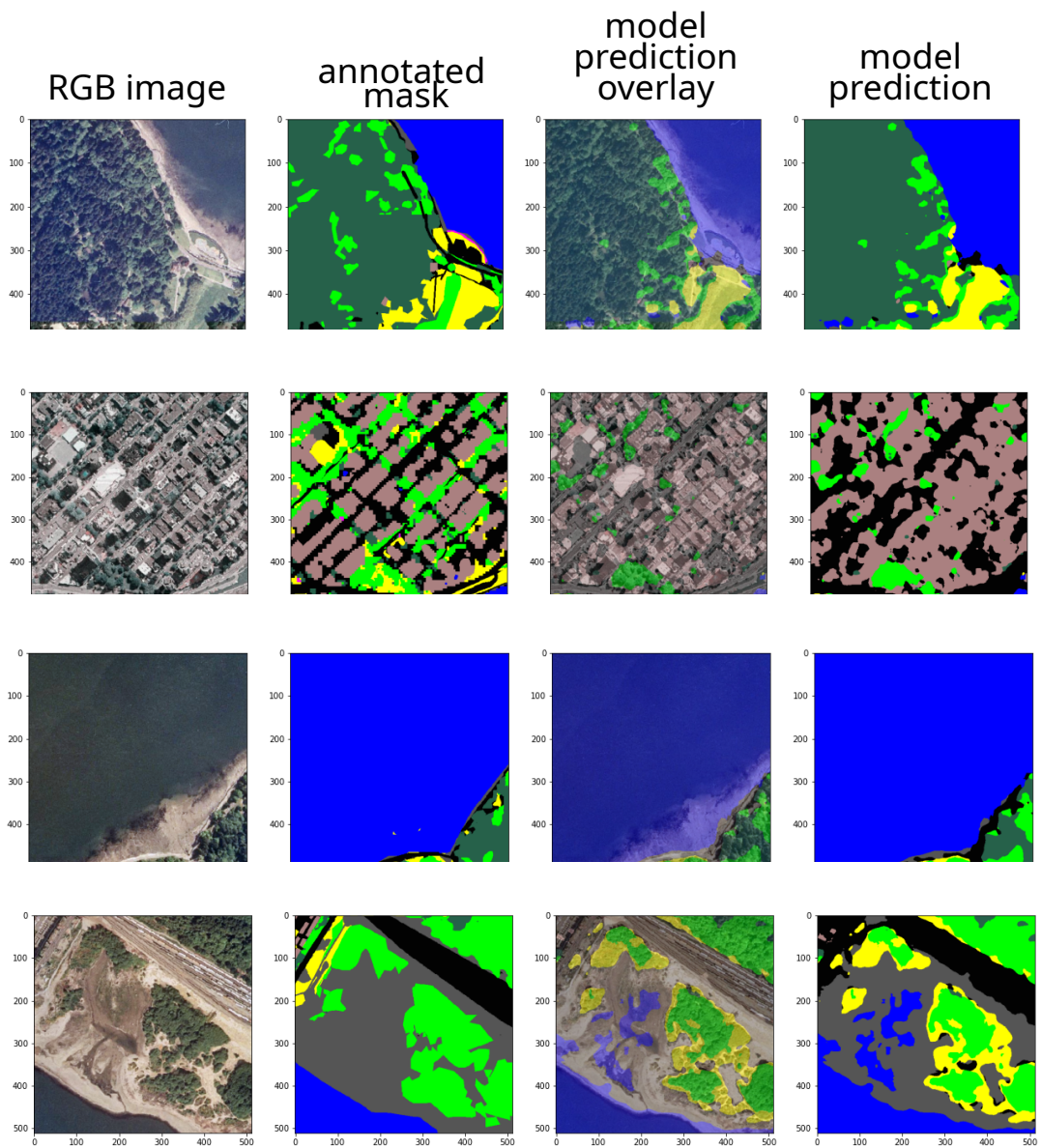


Figure S6: Map of Metro Vancouver, with black bounding box showing extent of training data. Basemap © OpenStreetMap.



-  water
-  buildings
-  pavement
-  barren
-  grass (including herb)
-  shrub
-  conifer
-  decidous

Figure S7: Historical RGB imagery tiles and land cover predictions.