



Catchment-scale benefits (flow reduction) and dis-benefits (debris blocking risk) of green-grey infrastructure in urban rivers

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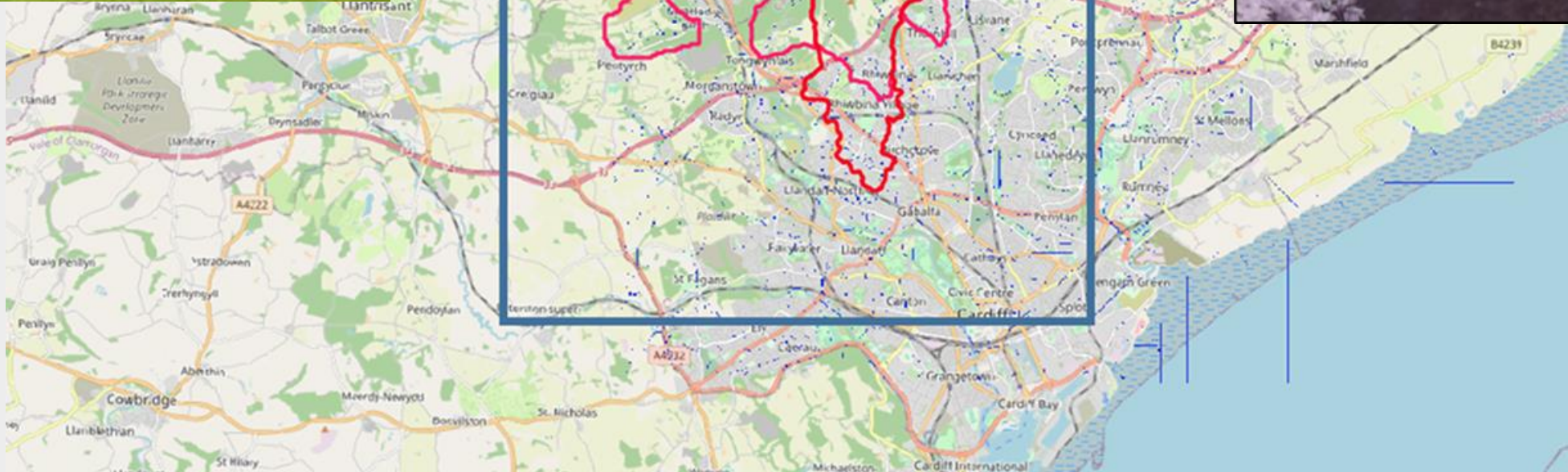
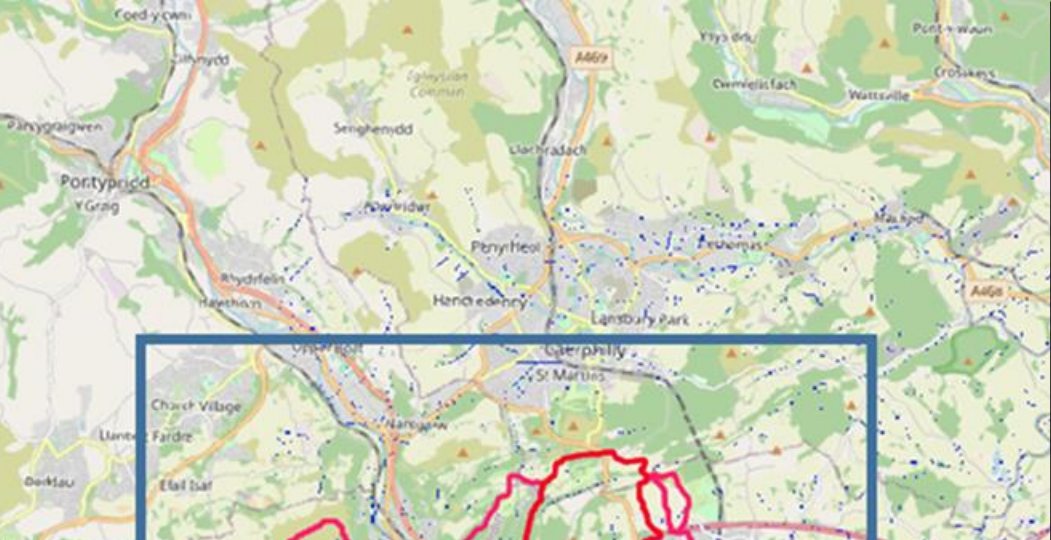
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Benefits and dis-benefits



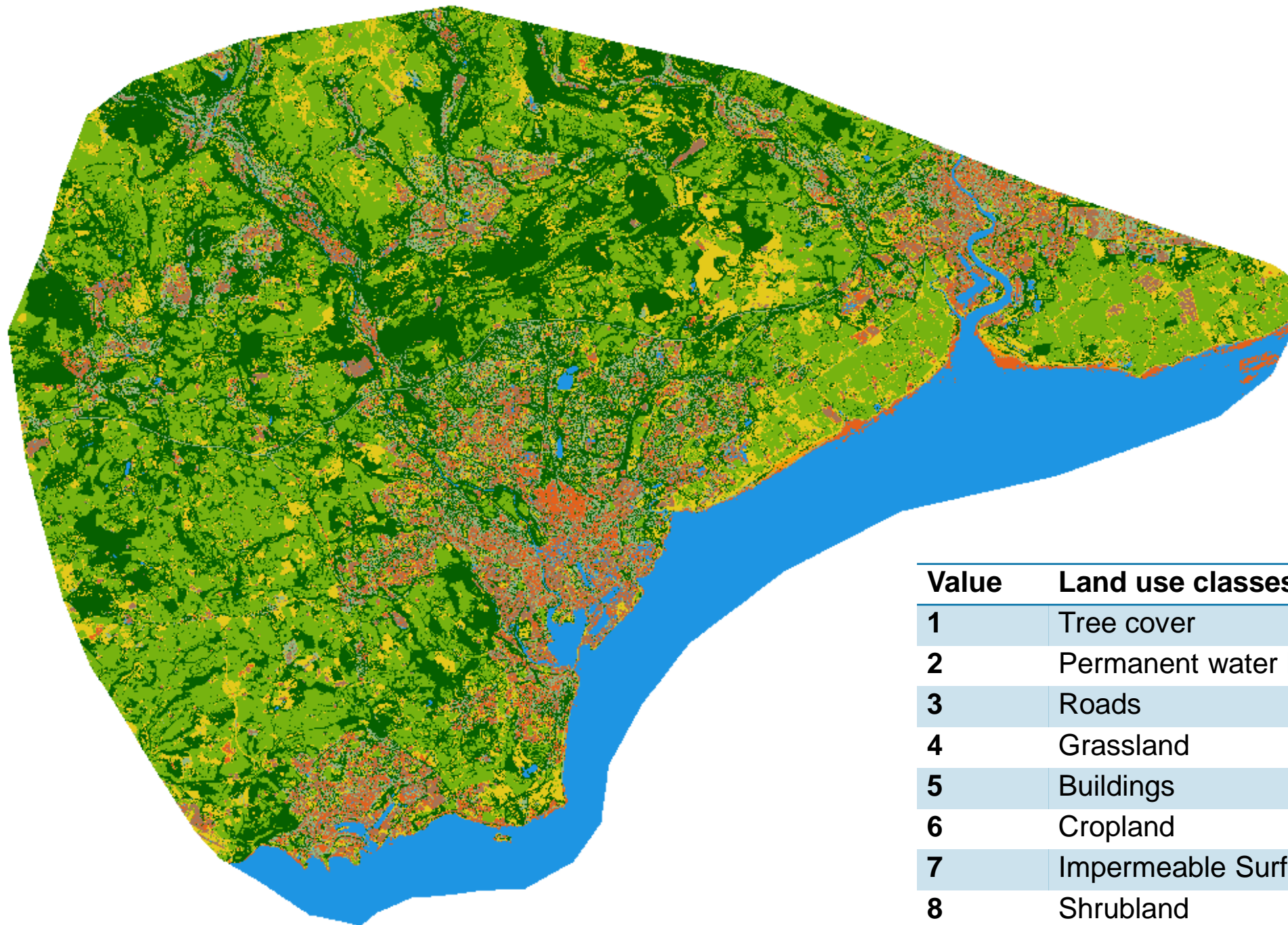
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Project Aims



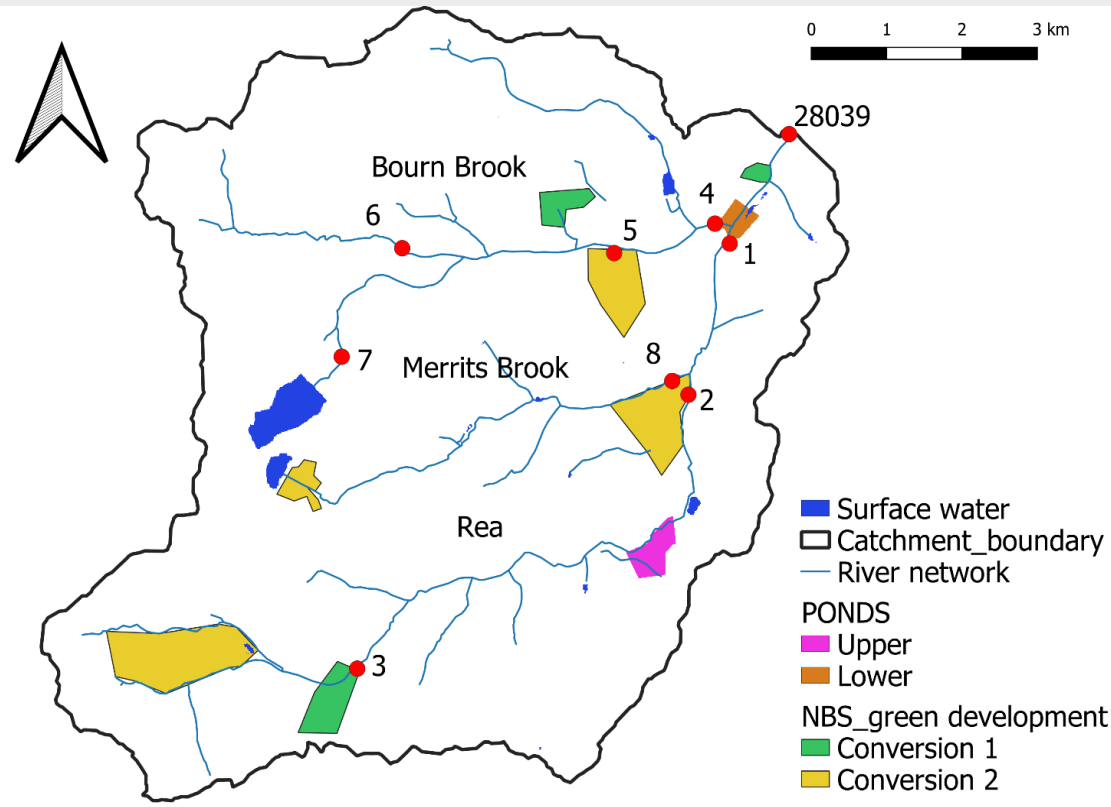
- » **O1:** Using existing GIS datasets to **map green and grey land-use across urban river catchments** draining to locations in Cardiff city where trash screens have been installed.
- » **O2:** Create new land use scenarios to **evaluate the hydrological benefit** (%change in flow) provided by existing GI. Parameterise and run the ANaRM model to evaluate those scenarios.
- » **O3:** Develop machine learning algorithm **to identify instances of debris blocking** of trash screens based on CCTV images and existing water level recordings.
- » **O4:** Use logistic regression to assess the relative contribution (statistical significance) of grey and green land-use characteristics on **the risk of blocking** across sites.



Value	Land use classes
1	Tree cover
2	Permanent water bodies
3	Roads
4	Grassland
5	Buildings
6	Cropland
7	Impermeable Surface
8	Shrubland

Adapted Nature-based solutions Rational Method (ANaRM)

Calibrated and validated in Birmingham, UK



» Rational method – urban, small catchment suitable – adapted to include *FARL*

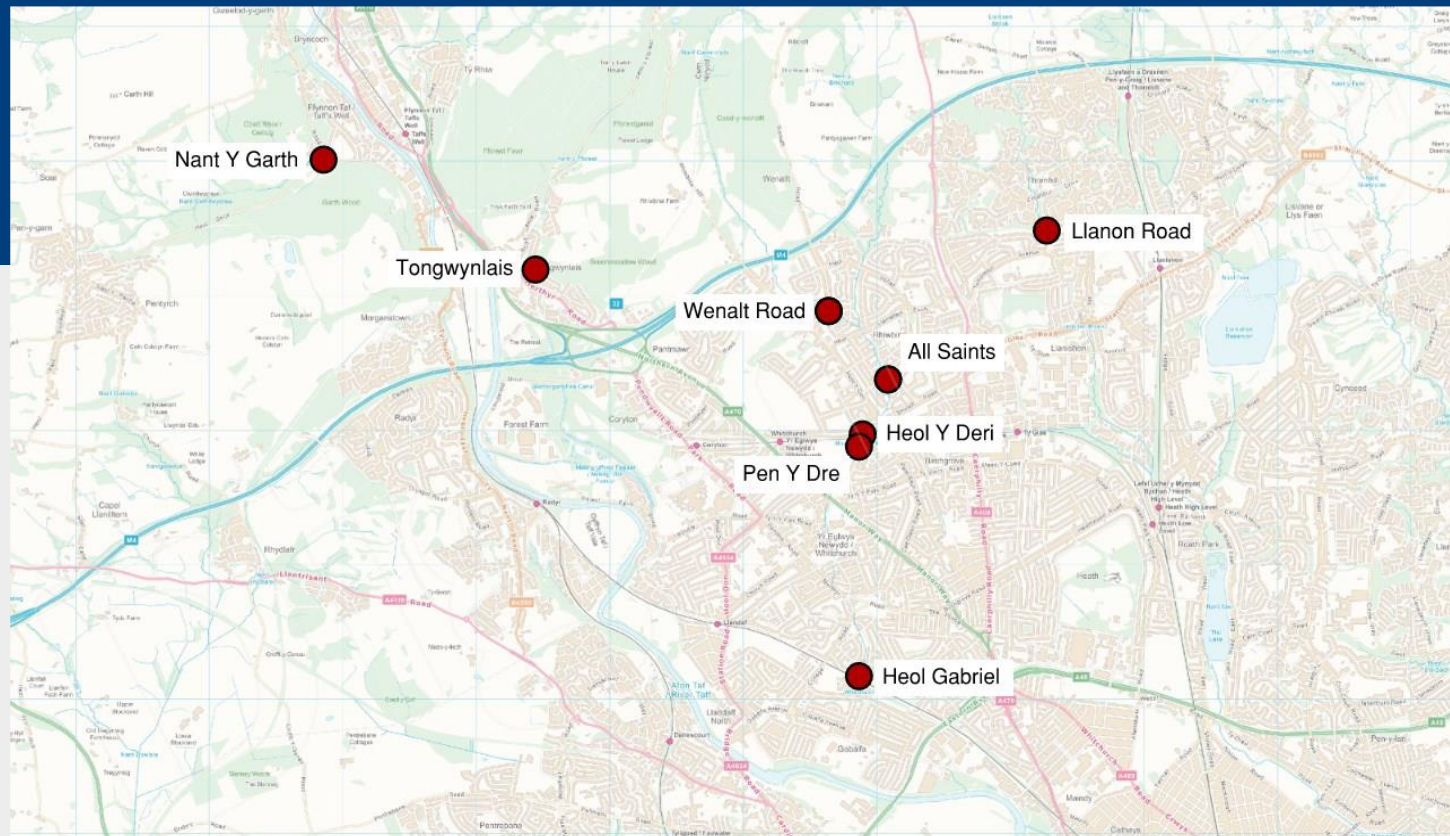
$Q_p = 0.278 C_i A \times FARL^{3.445}$ – gridded peak flow for any location in the catchment (m^3/s)

» $C = 1.3 \times C_v$ (*volumetric runoff coefficient*) – 0-1, depends on land cover type

» i = average rainfall intensity (mm/hr) > taken from FEH22 rainfall model

» A = flow accumulation raster area (km^2)

» *FARL* – flood attenuation from rivers/lakes (FEH catchment descriptor)



Nant Y Garth



Tongwynlais



Rhiwbina Wenalt Rd



Rhiwbina All Saints



Heol Gabriel



Llanon Road



Rhiwbina Heol Y Deri



Rhiwbina Pen Y Dre



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In total, 793 images of the Tongwynlais site were manually screened and labelled.

- 80% of the data was used for training
- 20% for validation

Label images as either **blocked** or **unblocked**

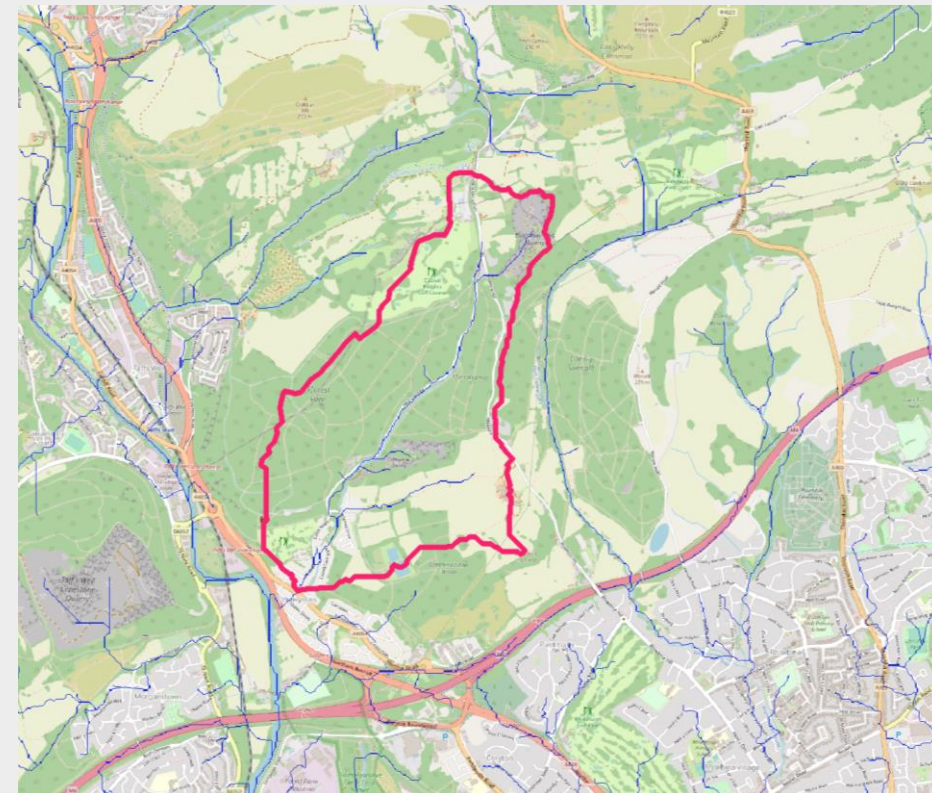


Image Subtraction



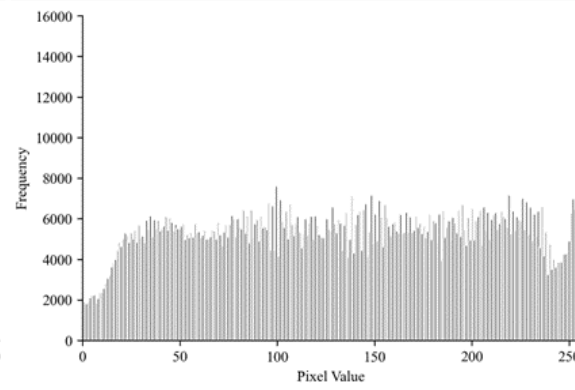
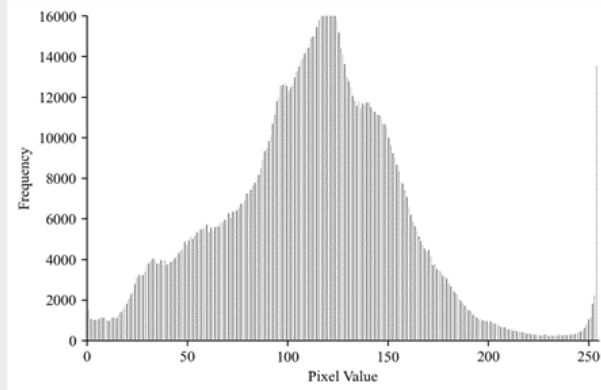
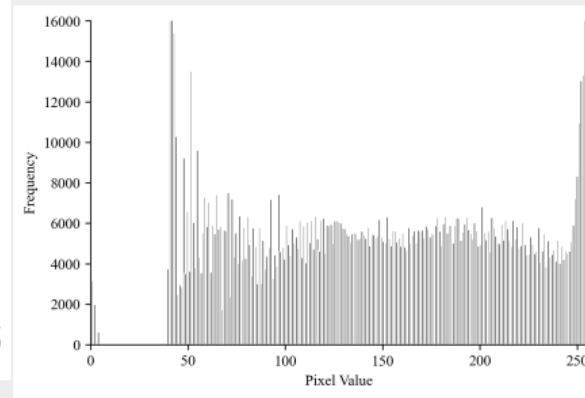
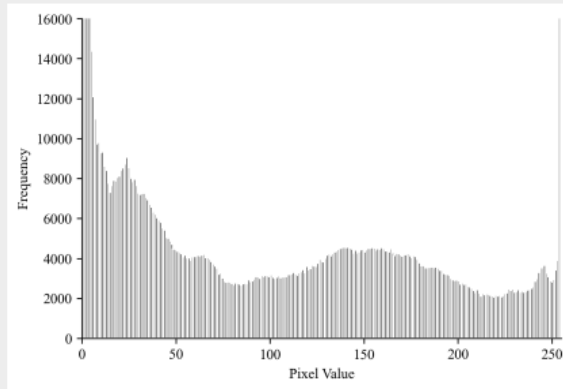
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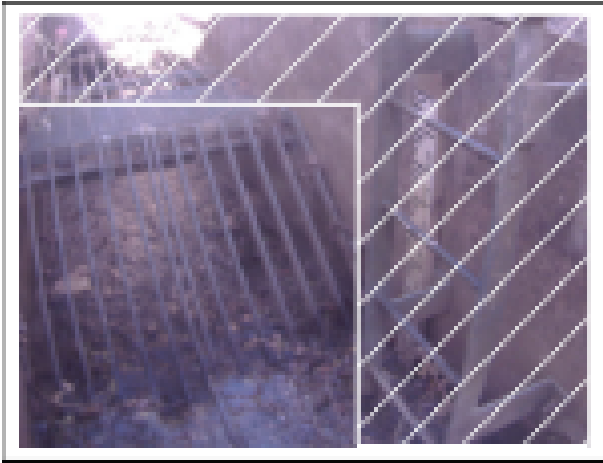


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Histogram equalisation





Input image cropped to
475x475 px



Image resized
by factor 0.1

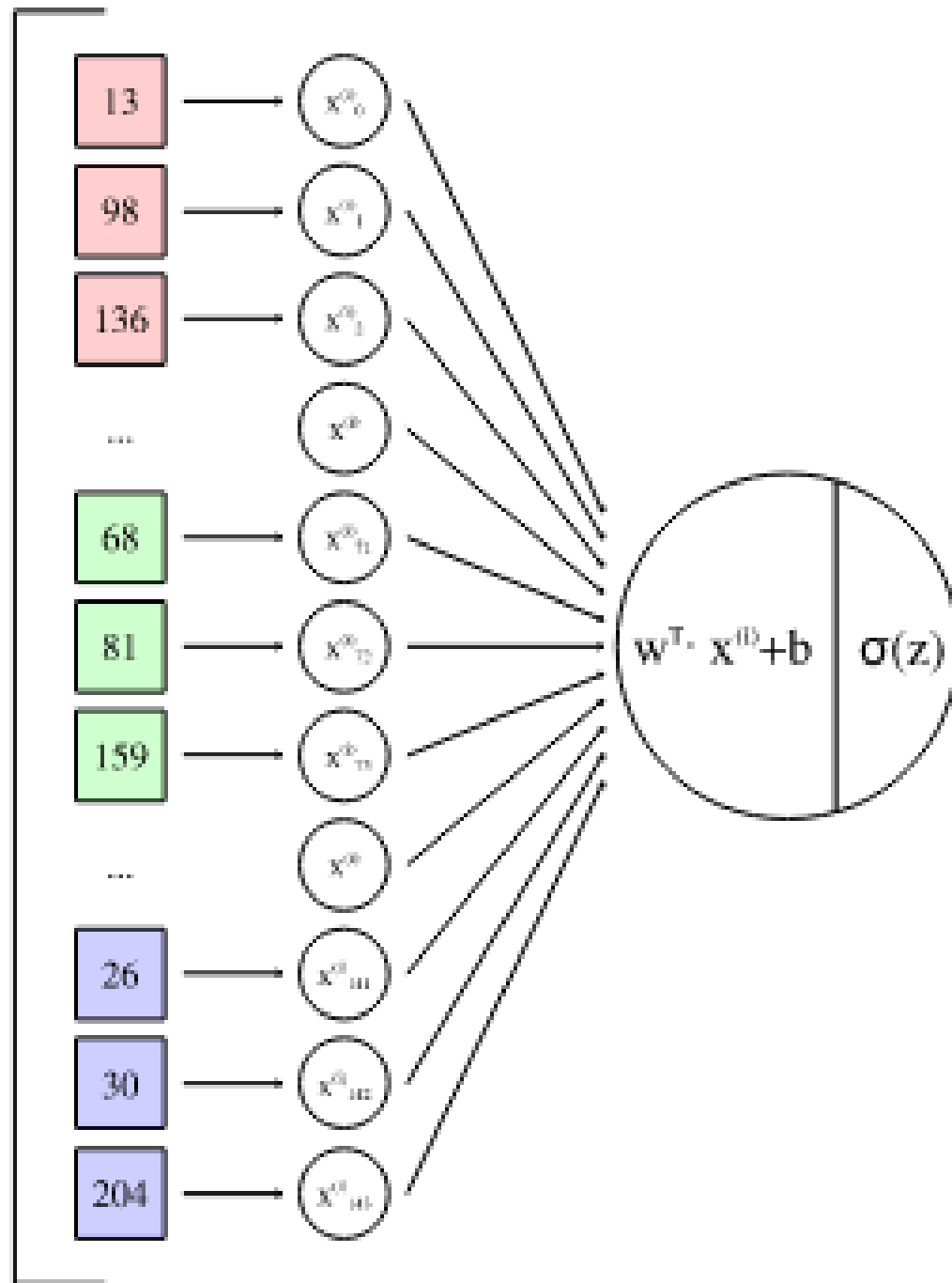


Table 4 Results of Logistic Regression and Time Taken to Train and Test Model with Dataset

Image Processing Type	Train Accuracy (%)	Test Accuracy (%)	Time Taken
M1: No additional processing	95.1	87.4	17.9 s
M2: Histogram Equalisation	97.1	91.0	19.7 s
M3: Canny Transformation	98.7	84.7	20.5 s
M4: Laplacian Transformation	100	89.2	23.4 s
M5: Laplacian Sharpened	100	90.1	25.9 s
M6: Image Subtraction	91.6	85.5	20.8 s
M7: Histogram Equalised Subtraction	96.2	87.4	22.4 s
M8: Canny Image Subtraction	98.0	85.6	26.8 s
M9: Laplacian Sharpened Subtraction	92.3	87.4	30.3 s
M10: No additional processing (0.5 scale)	100	91.0	325 s
M11: No additional processing (no scale)	100	90.1	1,470 s

Note. This Table contains results from cost function optimisations using 3000 model iterations and a learning rate of 0.01 using a Mid-range CPU (Intel Celeron N5095 @ 2.00GHz) with 16 G.B. RAM.

Next steps



- » Set-up ANARM model for Cardiff and trash screen catchments
- » Relate blocking frequency to land-use characteristics
- » Define future land-use scenarios
- » Investigate impact of land-use change -> Reduced flood peaks vs increased blocking frequency