

Automatic Detection of Ports For Map Generalisation

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**Automated generalisation for on-demand
mapping, Helsinki, Finland, 14 June, 2016**



Overview

- Ambition – to automatically detect higher order objects
 - Such as ‘farming region’, ‘mountainous regions’, ‘suburbs’, ‘ports’...
 - Derived from open source data
- OSM – entities recorded at a specific conceptual scale
 - → Therefore ‘entity constrained’
 - → Higher order objects not explicitly modelled
- Methodology
 - Detection of clusters of attributes of higher order objects
 - Use machine learning and neural nets to search for prototypical higher order objects



Derivation of very small scale mapping from large scale open source data



Major UK Ports



OSM



OSM

- Physical features represented using tags attached to basic data structures (nodes, ways, relations).
- Each tag describes a geographic attribute of the feature being shown by that specific node, way or relation.
- ‘Free tagging’ allows the map to include an unlimited number of attributes describing each feature
- Scale dependent nature of entities



How is 'Port' defined within OSM?

1.13 Leisure

1.14 Man Made

1.15 Military

1.6 Building

1.6.1 Accommodation

1.6.2 Commercial

1.6.3 Civic/Amenity

1.6.4 Other Buildings

1.6.5 Additional Attributes

1.16 Natural

1.16.1 Veg/ surface related

1.16.2 Water related

1.16.3 Landform related

1.18 Places

1.18.1 Administratively declared place

1.18.2 Populated settlements, urban

1.18.3 Pop. settlements, urban & rural

1.18.4 Other places

1.18.5 Additional attributes

1.26 Waterway

1.26.1 Natural watercourses

1.26.2 Man made waterways

1.26.3 Facilities



An Ontological Perspective

Port

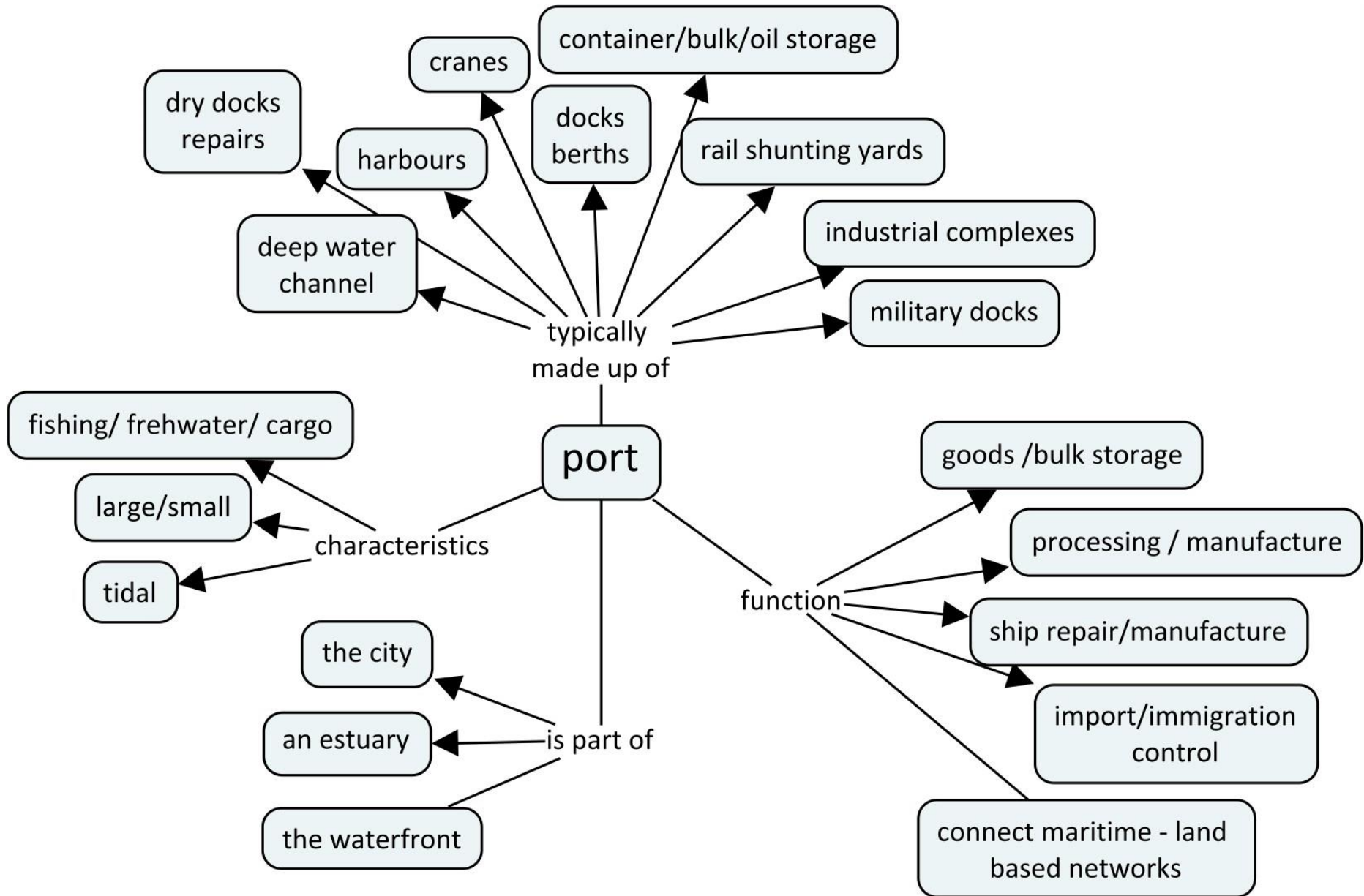
‘a location on a coast or shore containing one or more harbours where ships can dock and transfer people or cargo to or from land’

Ontology

An ontology is a shared conceptual description that facilitates machine-interpretable definitions of basic concepts in a domain and the relations among them

The role of the ontology: competency questions



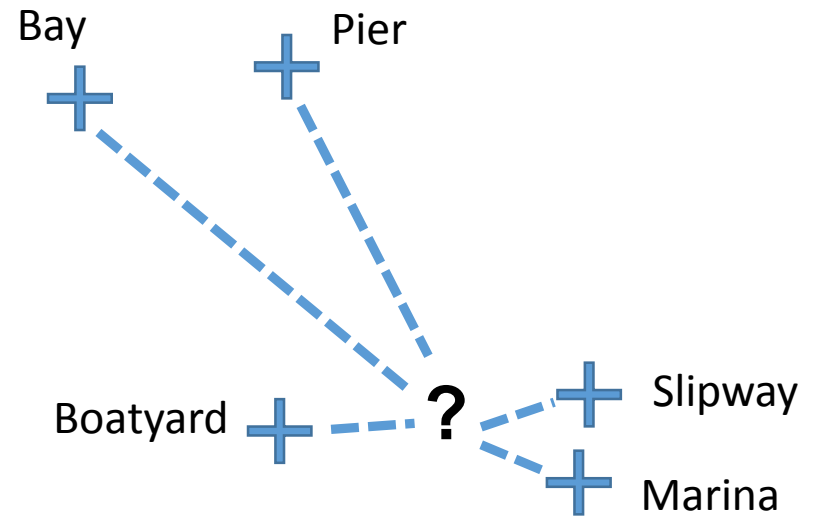
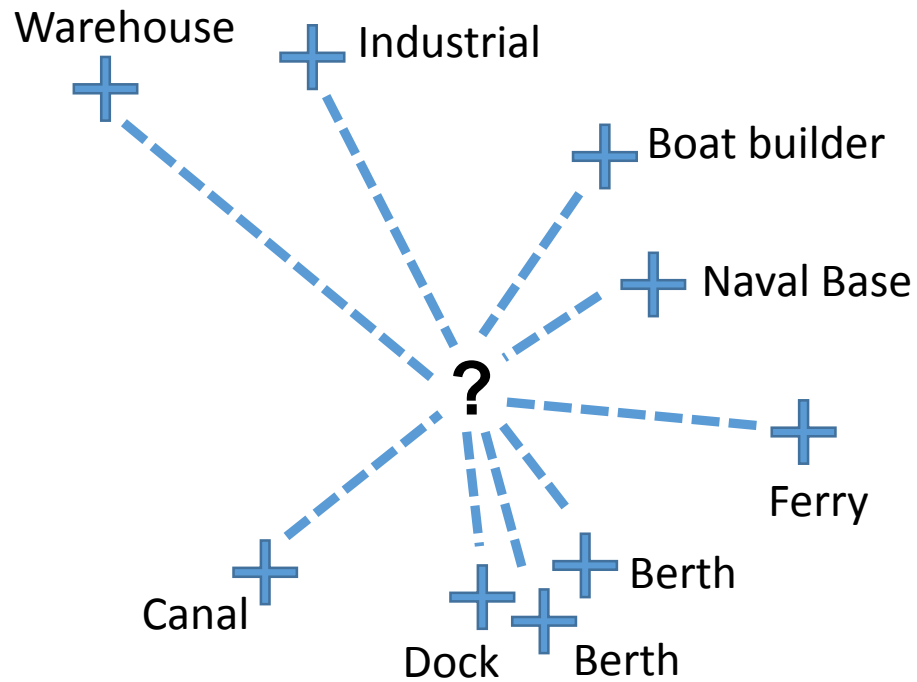


Variable	Contains the word
Amenity	boat
Amenity	ferry
Boundary	maritime
Building	industrial
Building	warehouse
Building	train station
Craft	boat builder
Craft	sail
Emergency	life
Historic	ship
Landuse	commercial
Landuse	industrial
Landuse	military
Landuse	port
leisure	marina
Leisure	slipway
Seamark	Berth
Seamark	Boatyard
Seamark	slipway

Variable	Contains the word
man-made	breakwater
man-made	crane
man-made	lighthouse
man-made	pier
man-made	silo
man-made	storage
military	naval base
natural	bay
natural coast	
power	plant
public transport	station
railway	rail
railway station	
service	siding
service	yard
route	ferry
waterway	canal
waterway	dock
waterway	boatyard



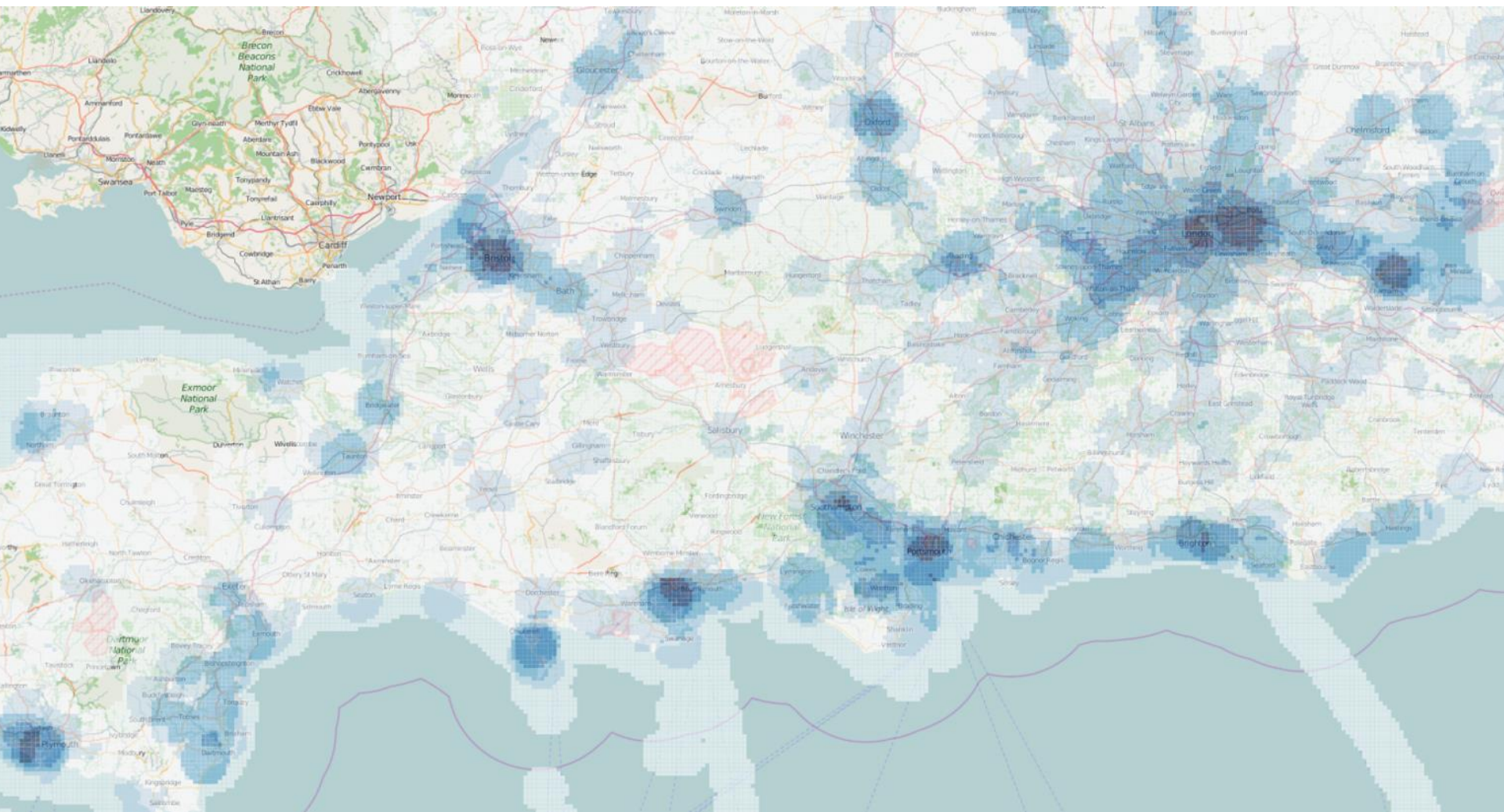
Major & Minor Ports?



Methodology

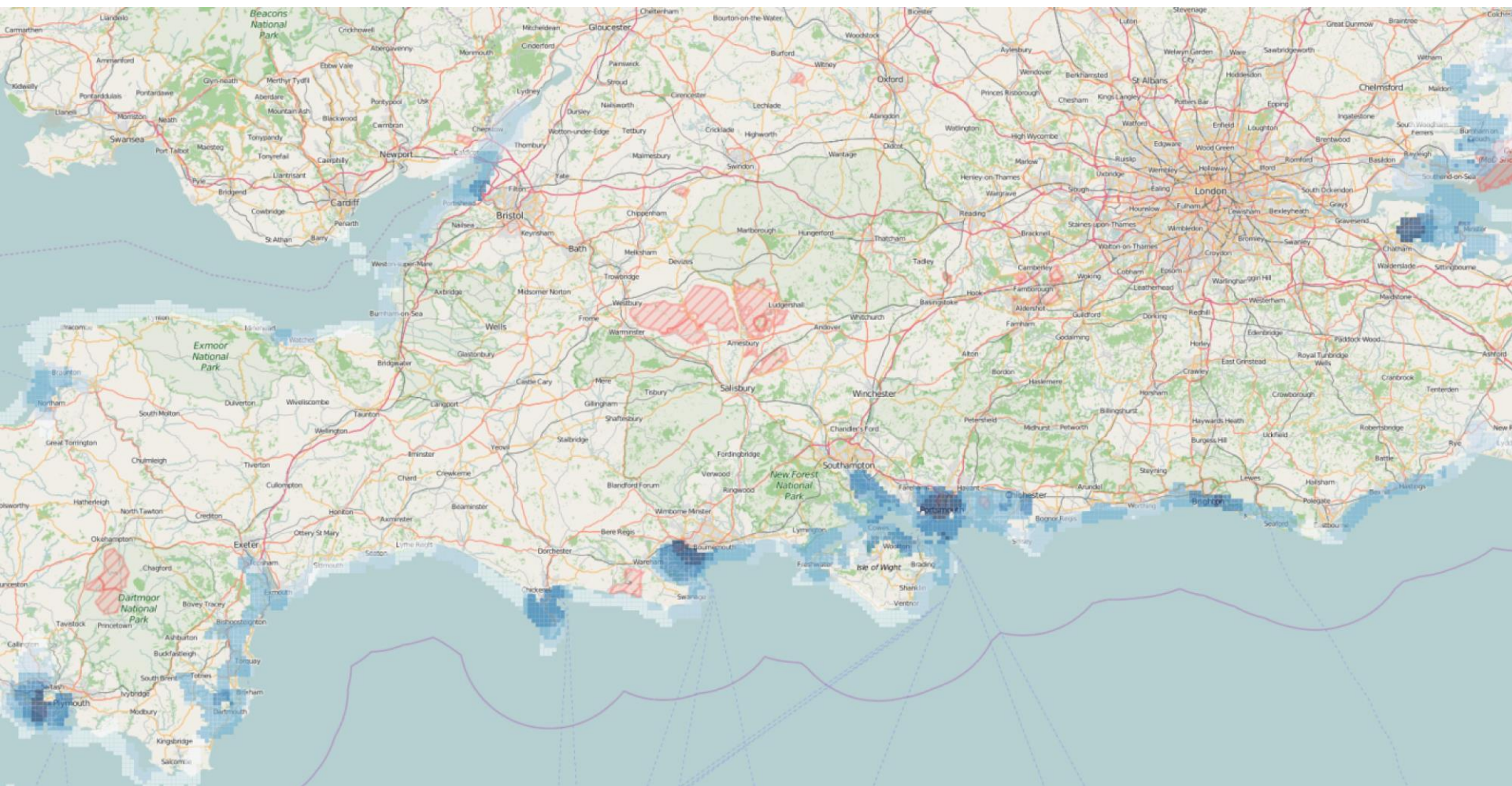
- Create UK Fishnet
 - 1km by 1km square
 - 960,000 cells
 - Coastline modelling: Boolean: within 1km of coastline
- For each cell
 - calculated distance to nearest item from a list of 38 different features... (eg railway siding, marina, pier)
 - If feature not found within 5km: 999 score awarded
 - Calculate the mean of the sum of distances





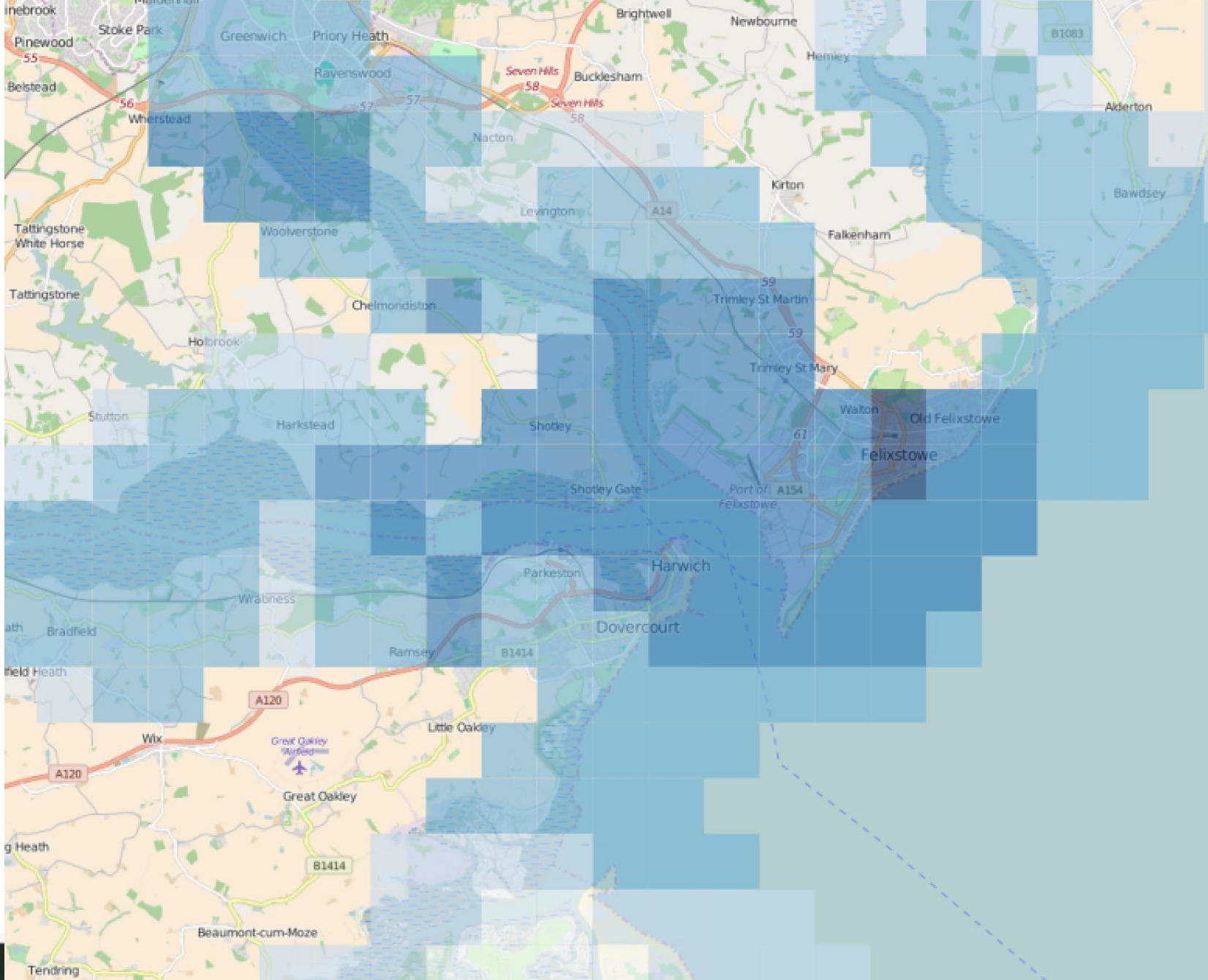
Inference: dark blue → clustering of Port sub-entities

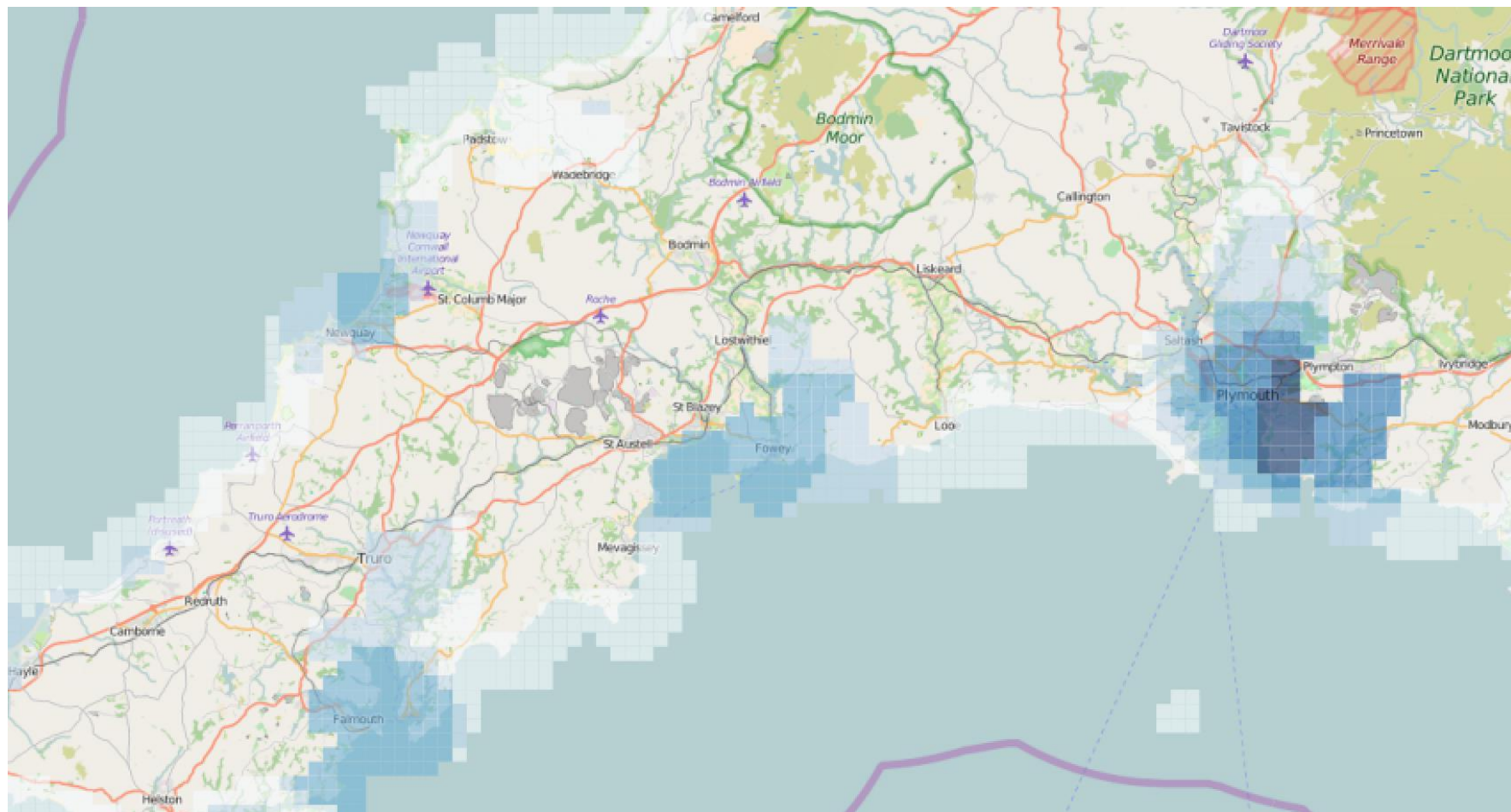


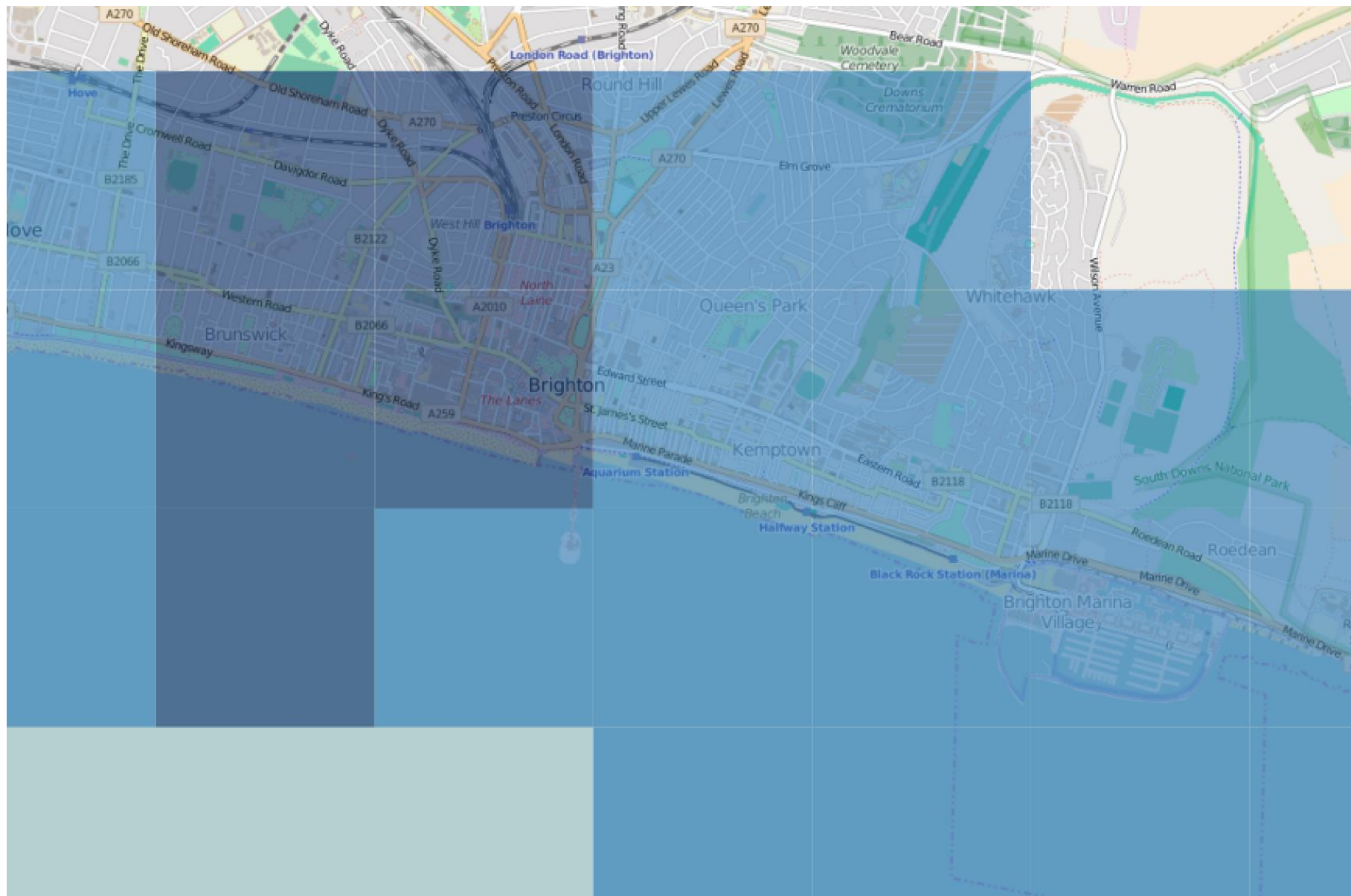


Inference: dark blue → clustering of Port sub-entities







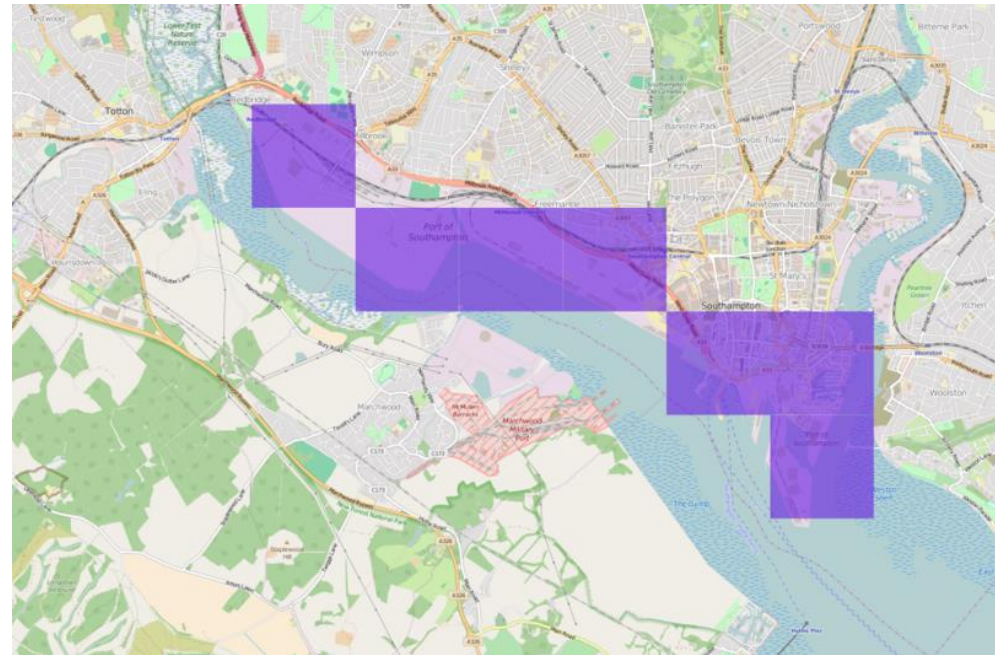
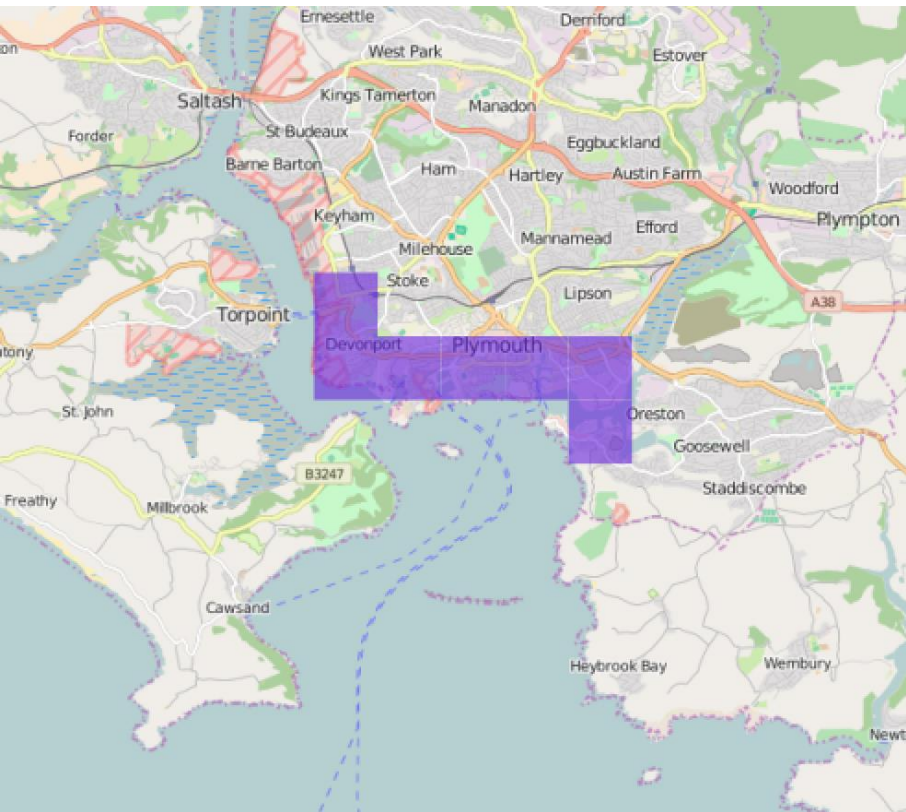


Machine Learning?

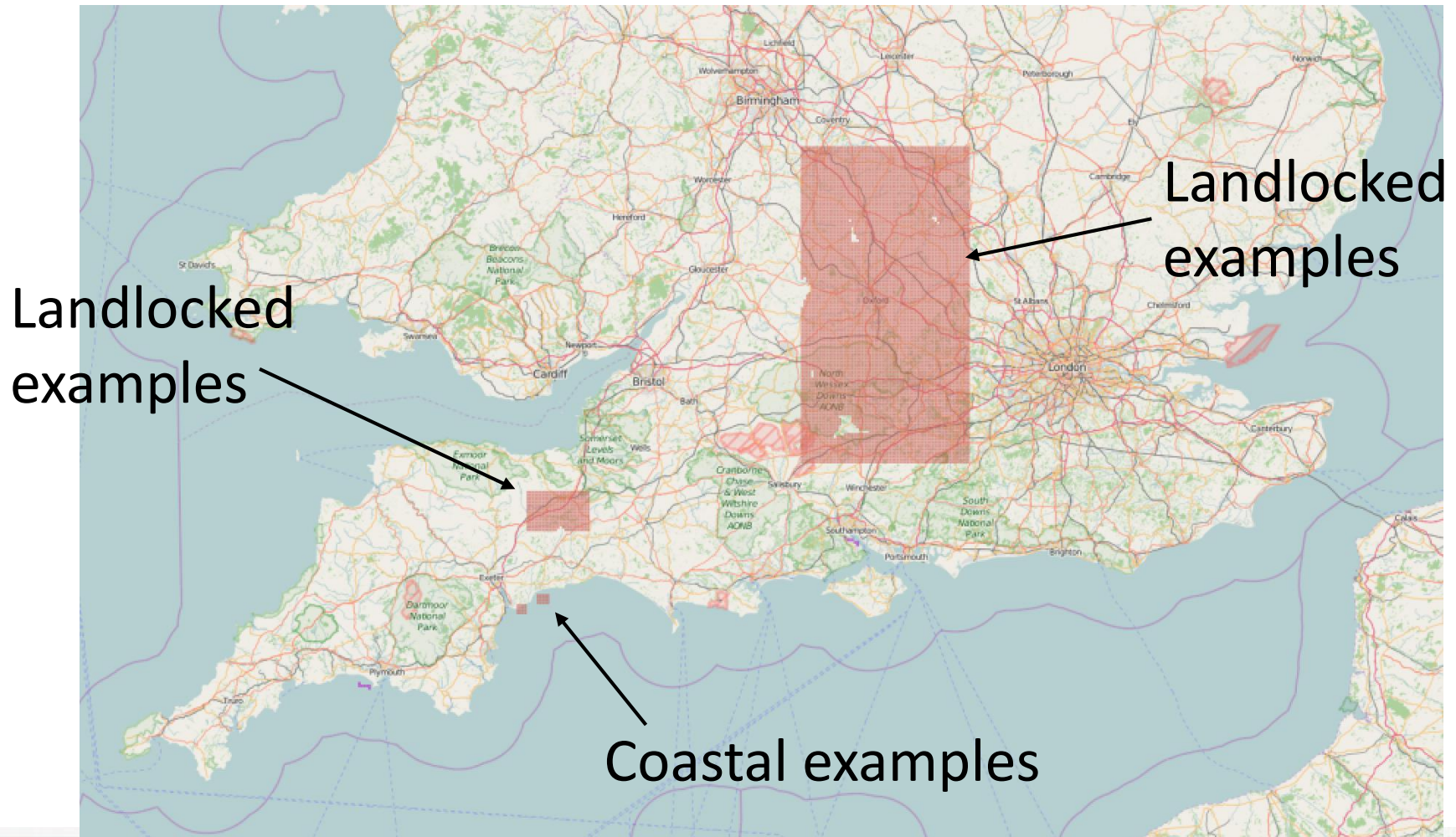
- evolved from the study of pattern recognition and computational learning theory in AI
- Field of study that gives computers the ability to learn without being explicitly programmed.
- 3 Approaches
 - Decision Trees
 - Bayes Classifier
 - Neural Net



Training set



Training set



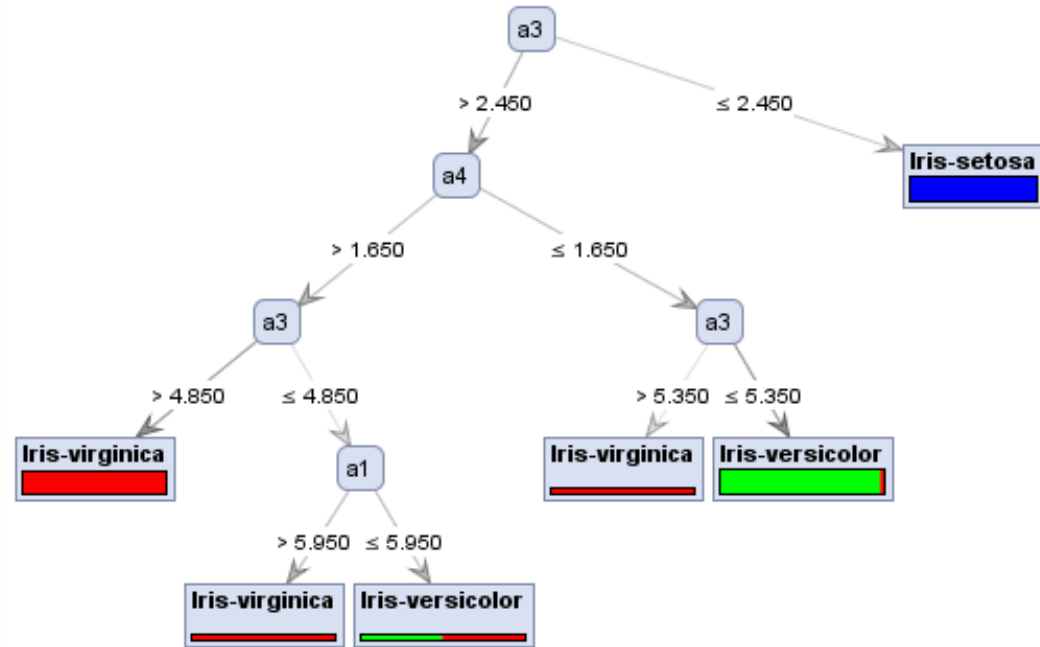
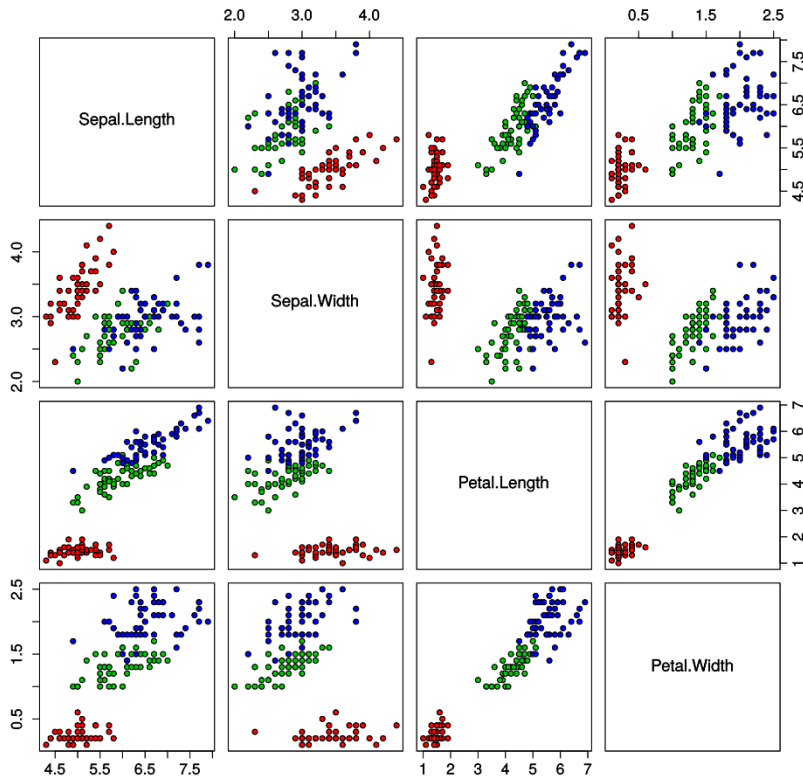
Decision Trees

- Uses a decision tree as a predictive model which maps observations about an item to conclusions about the item's target value
- Tree structure denotes decision pathways
 - Comprises decision nodes & leaf nodes
- Topmost decision node corresponds to best predictor – root node
- Handle both categorical & numerical data

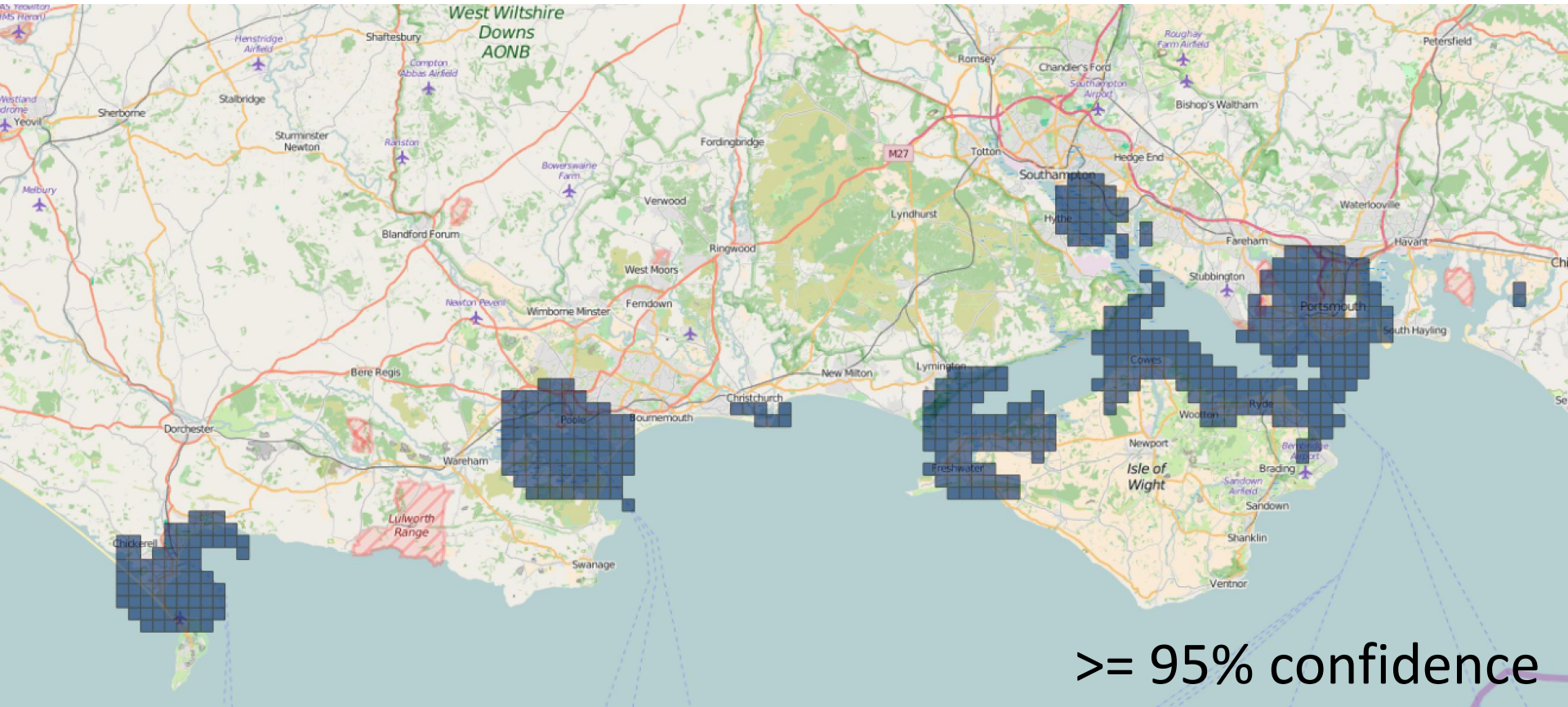


Decision tree: classification Iris Data

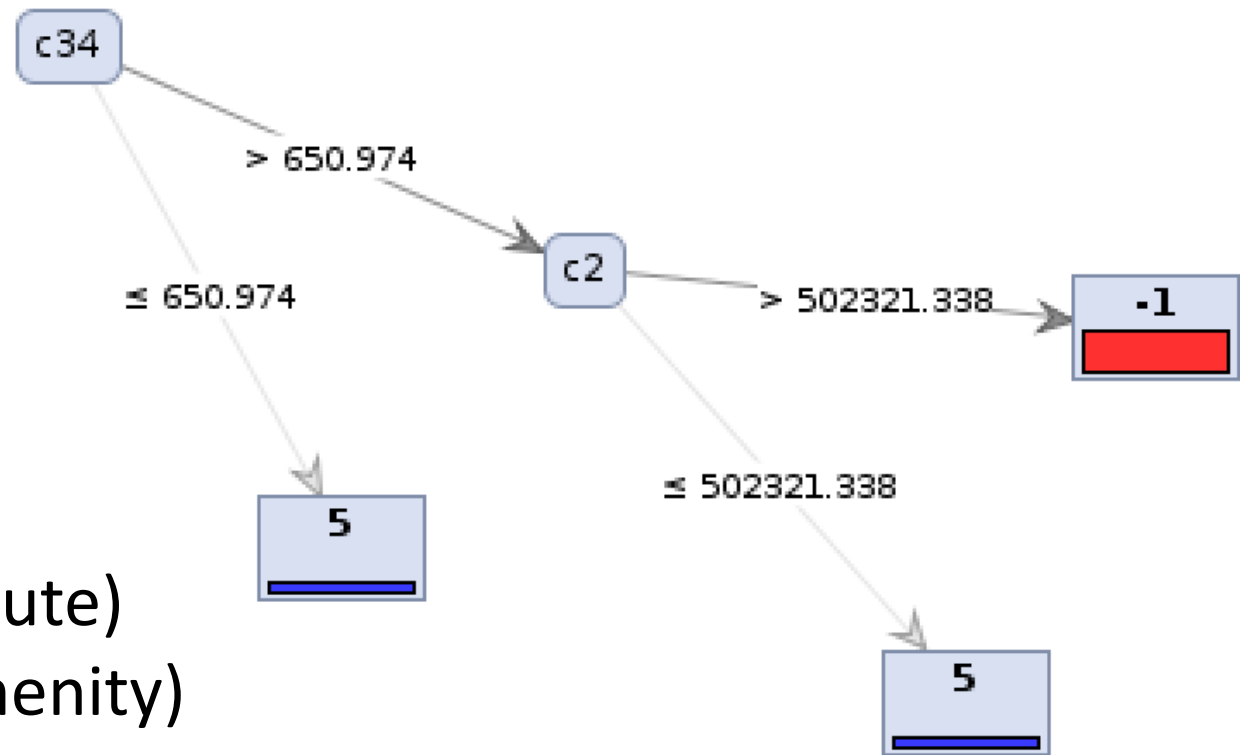
Iris Data (red=setosa, green=versicolor, blue=virginica)



Decision Tree



Decision Tree



C34 = FERRY (route)

C2 = FERRY (amenity)

5 = Port

-1 = Not a port

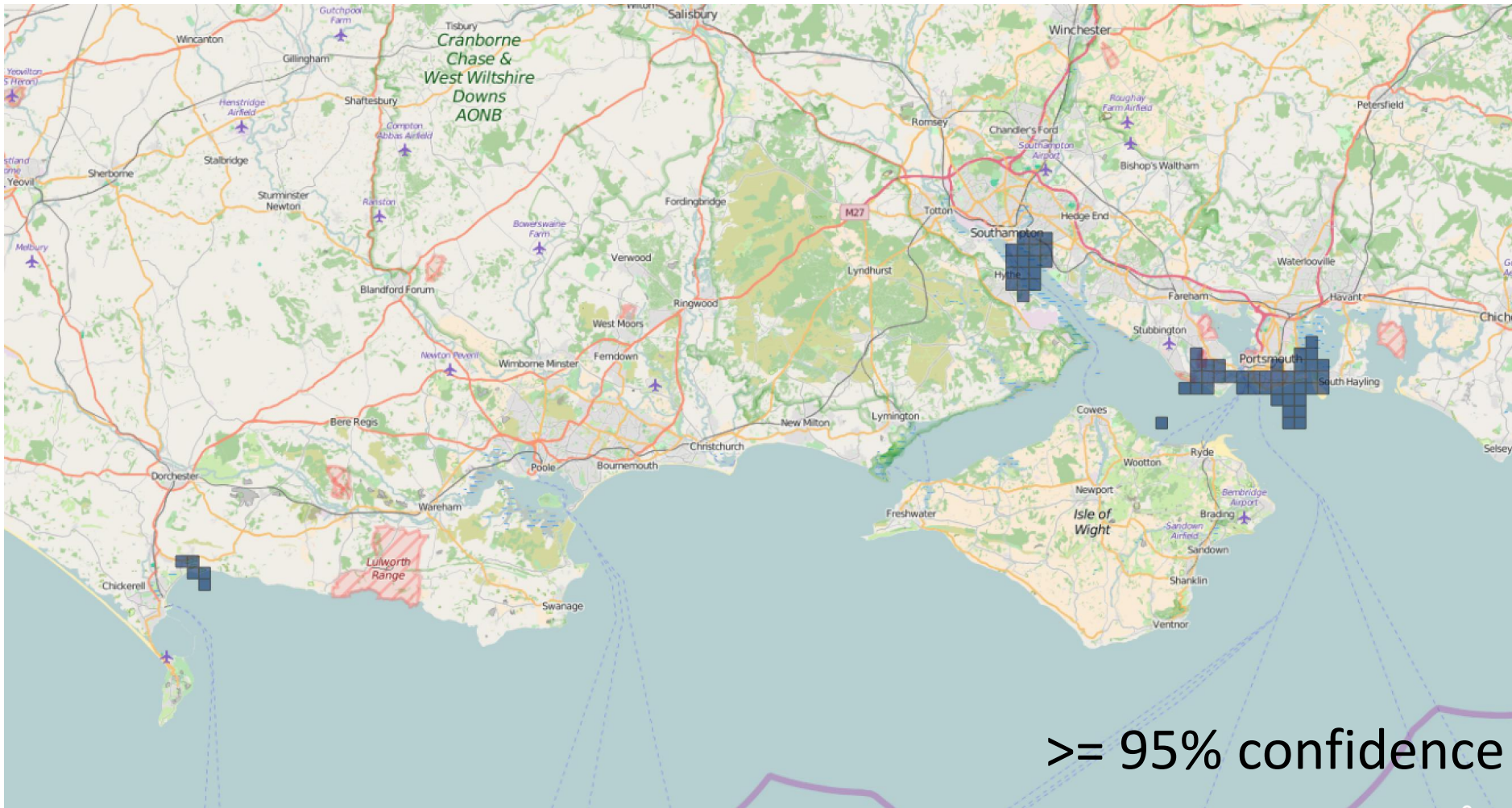


Baye's Classifier

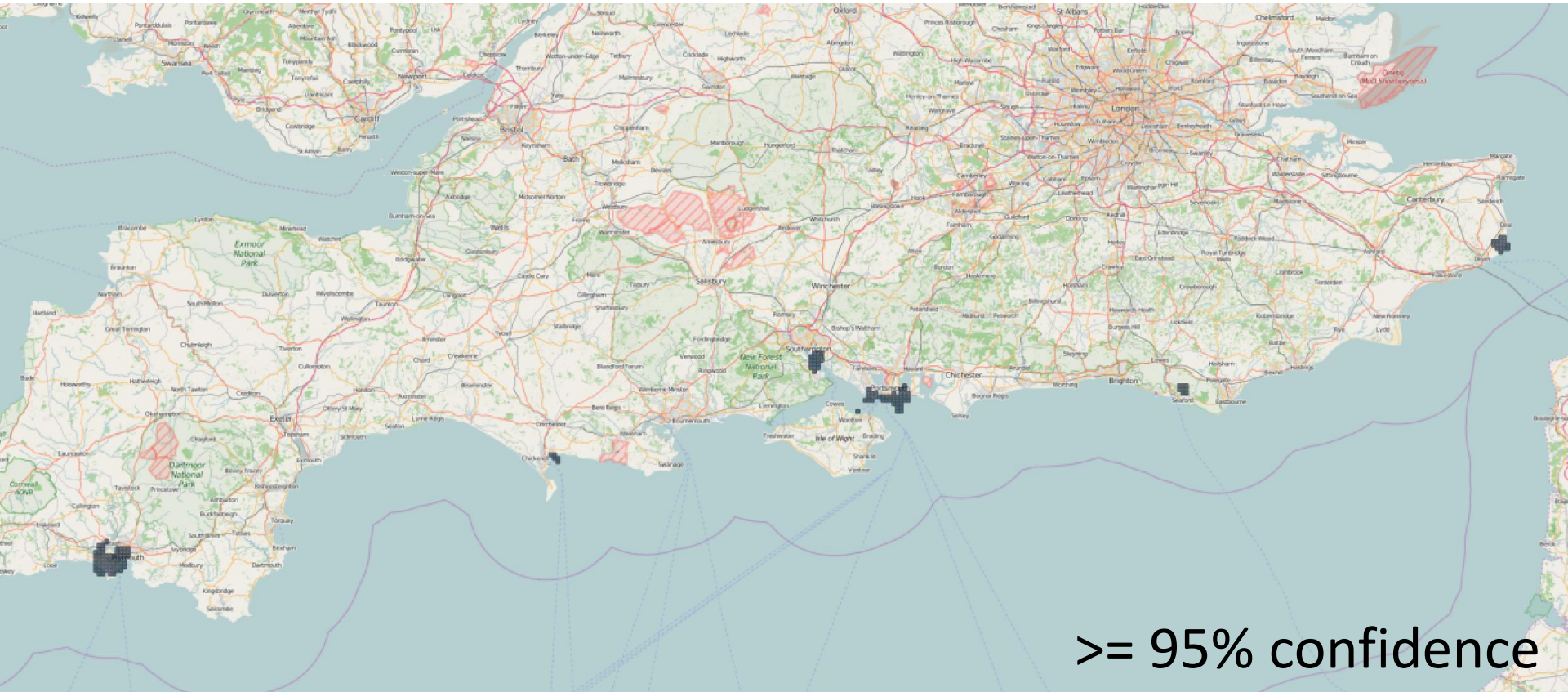
- is a probabilistic graphical model (a type of statistical model) that represents a set of random variables and their conditional dependencies via a directed acyclic graph (DAG).



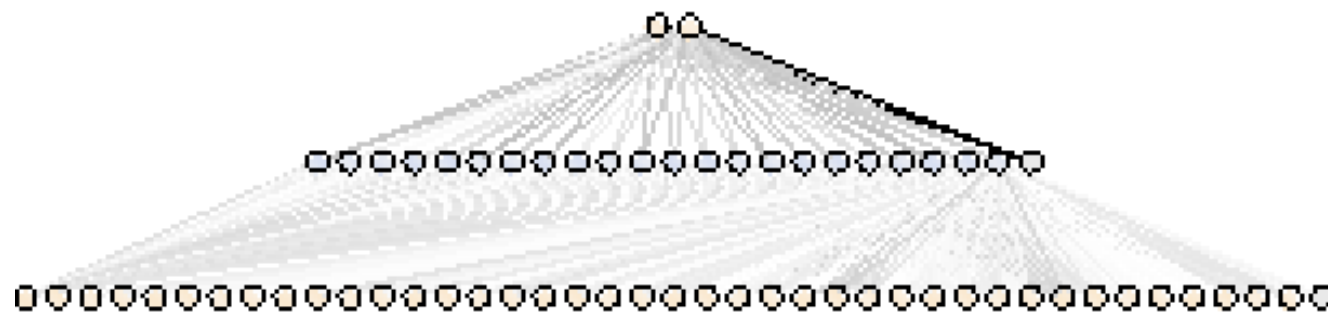
Bayes Classifier



Bayes Classifier



Artificial Neural Network



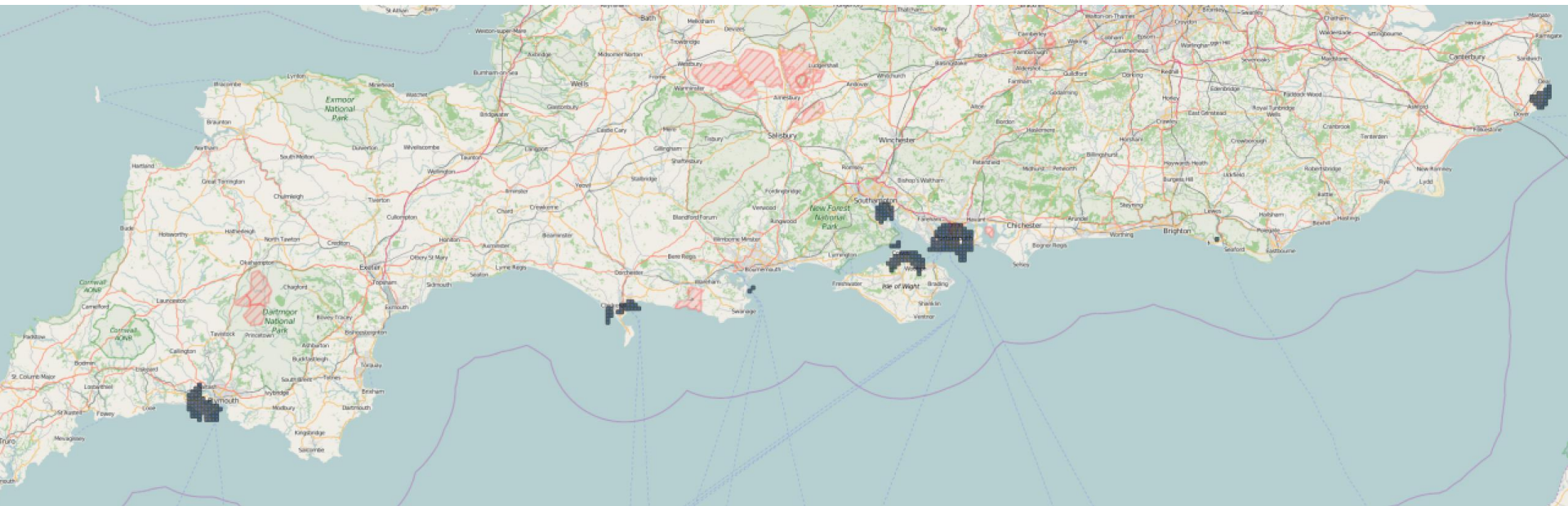
2 output classes
(port or not port)

1 hidden layer

38 inputs



Artificial Neural Network: outputs



$\geq 95\%$ confidence



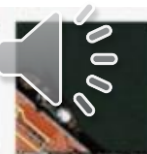
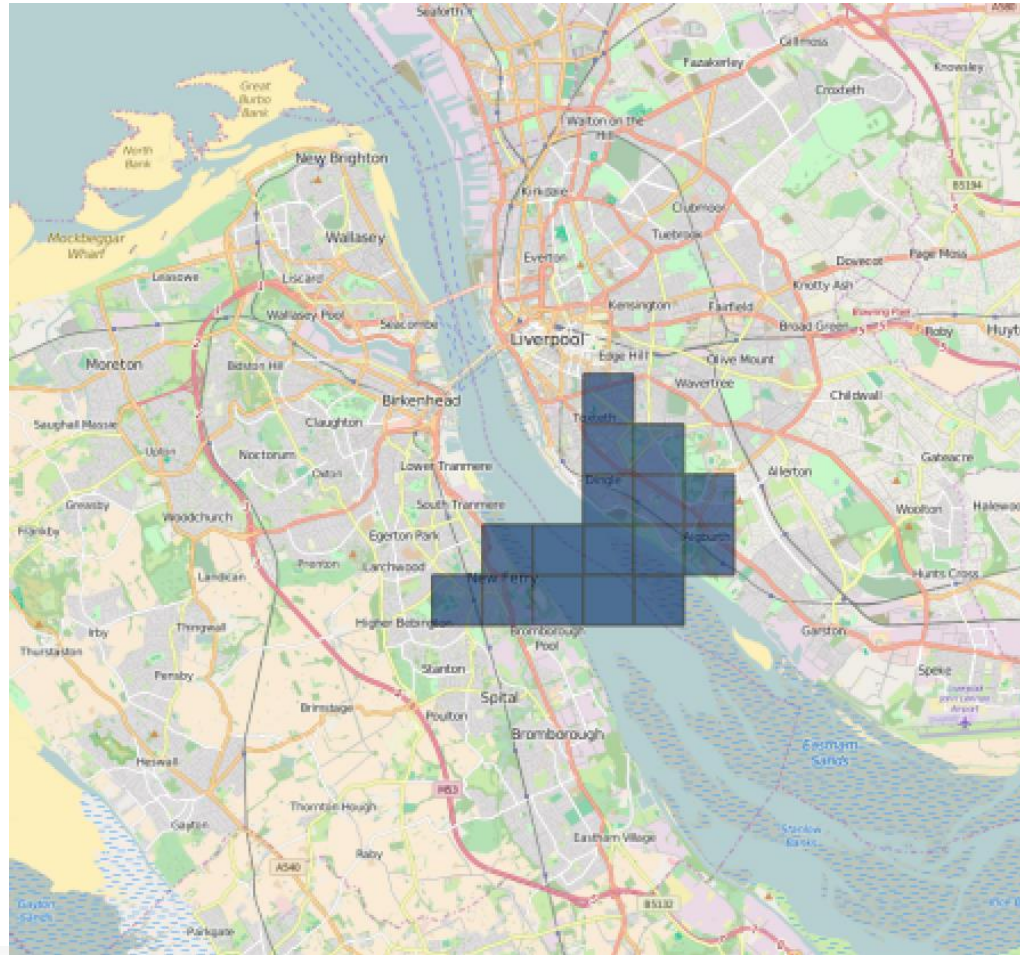
Artificial Neural Network: outputs



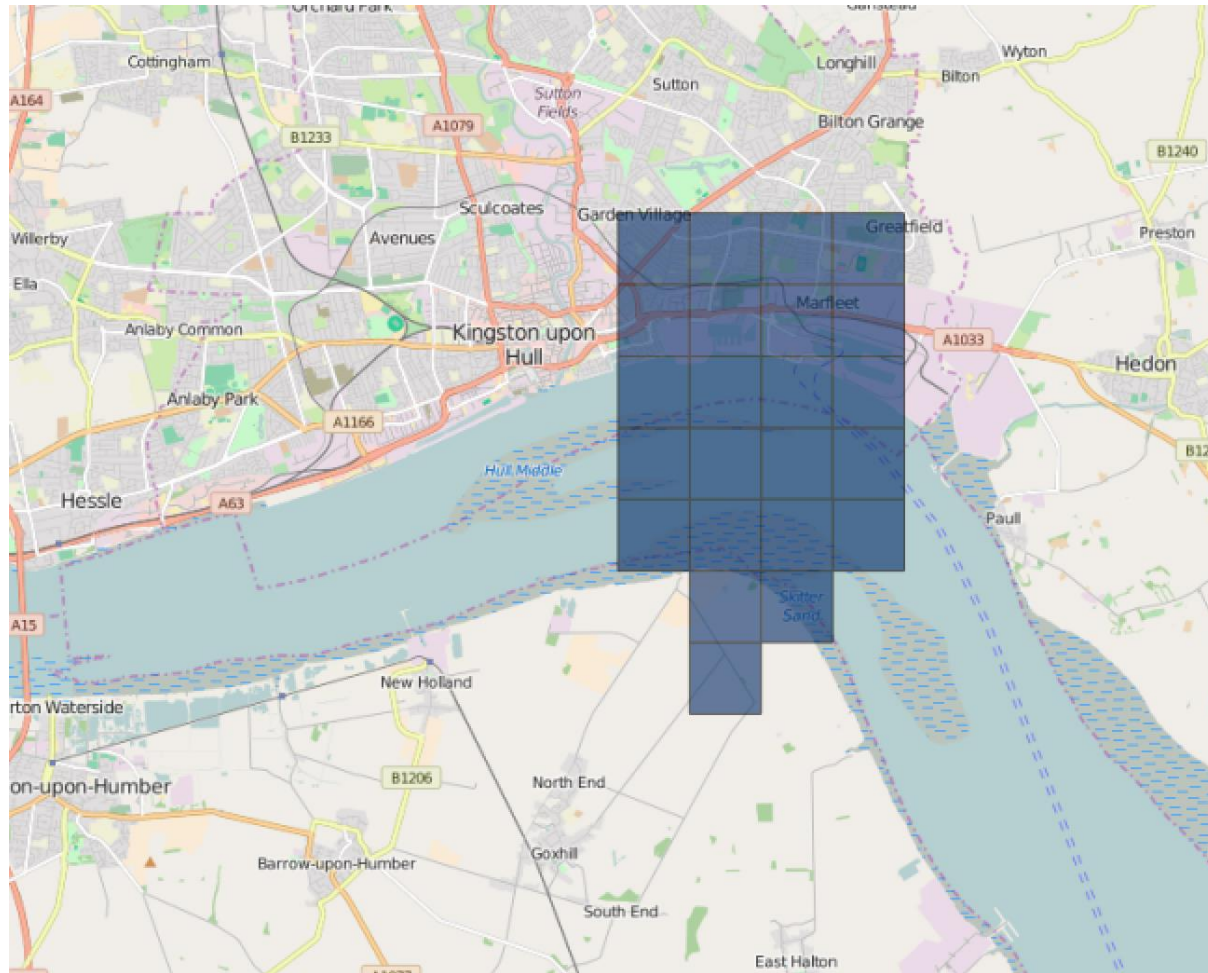
$\geq 95\%$ confidence



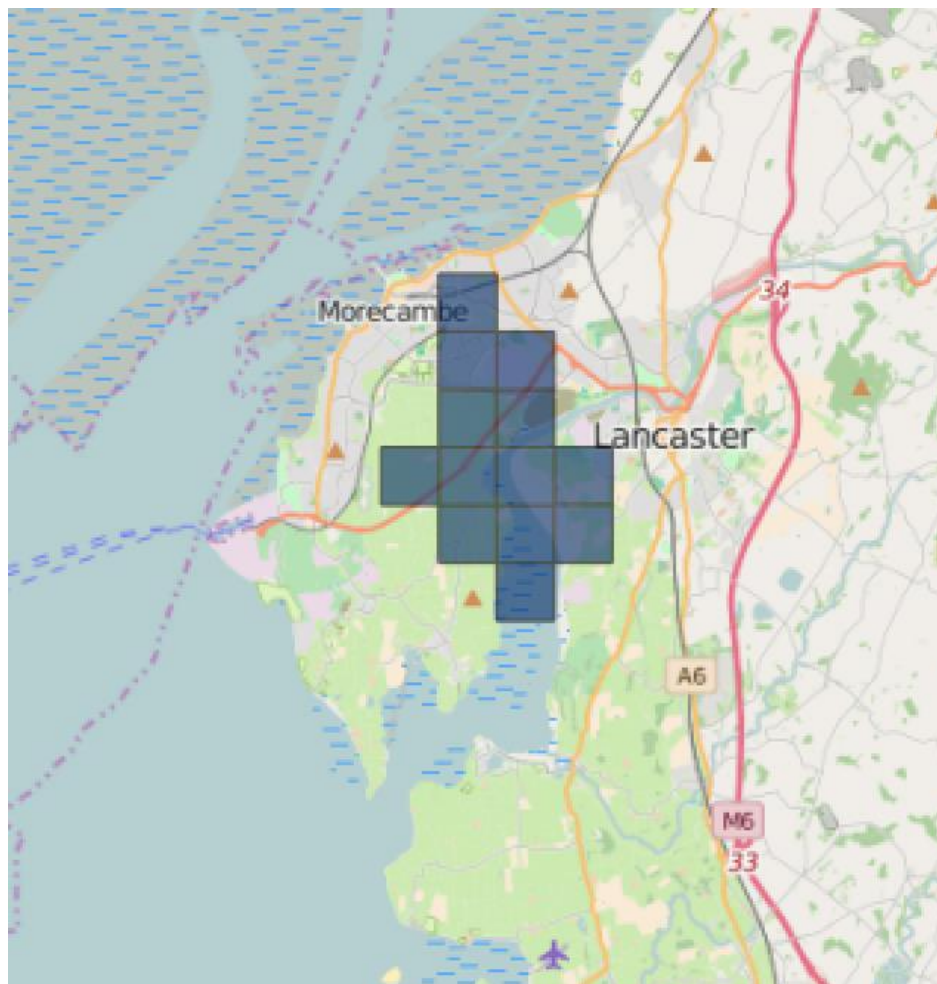
Artificial Neural Network: outputs



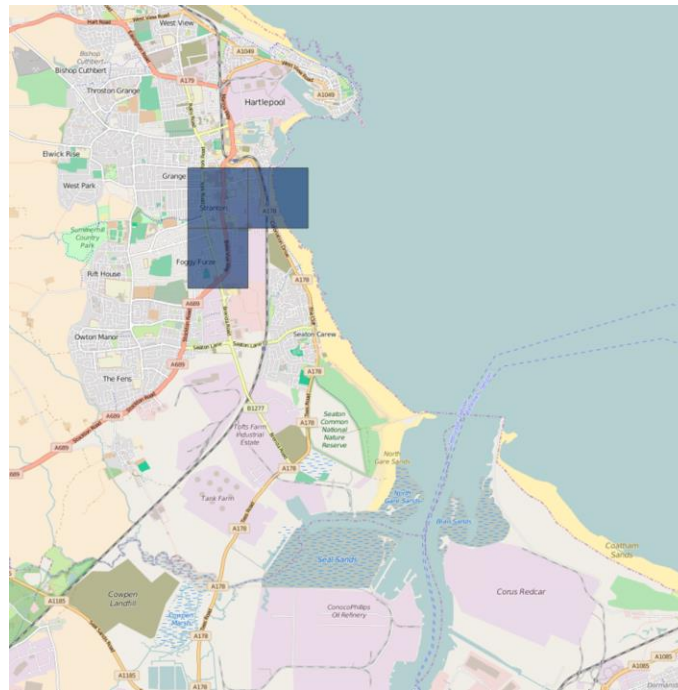
Artificial Neural Network: outputs



Artificial Neural Network: outputs



Artificial Neural Network: outputs



Evaluation

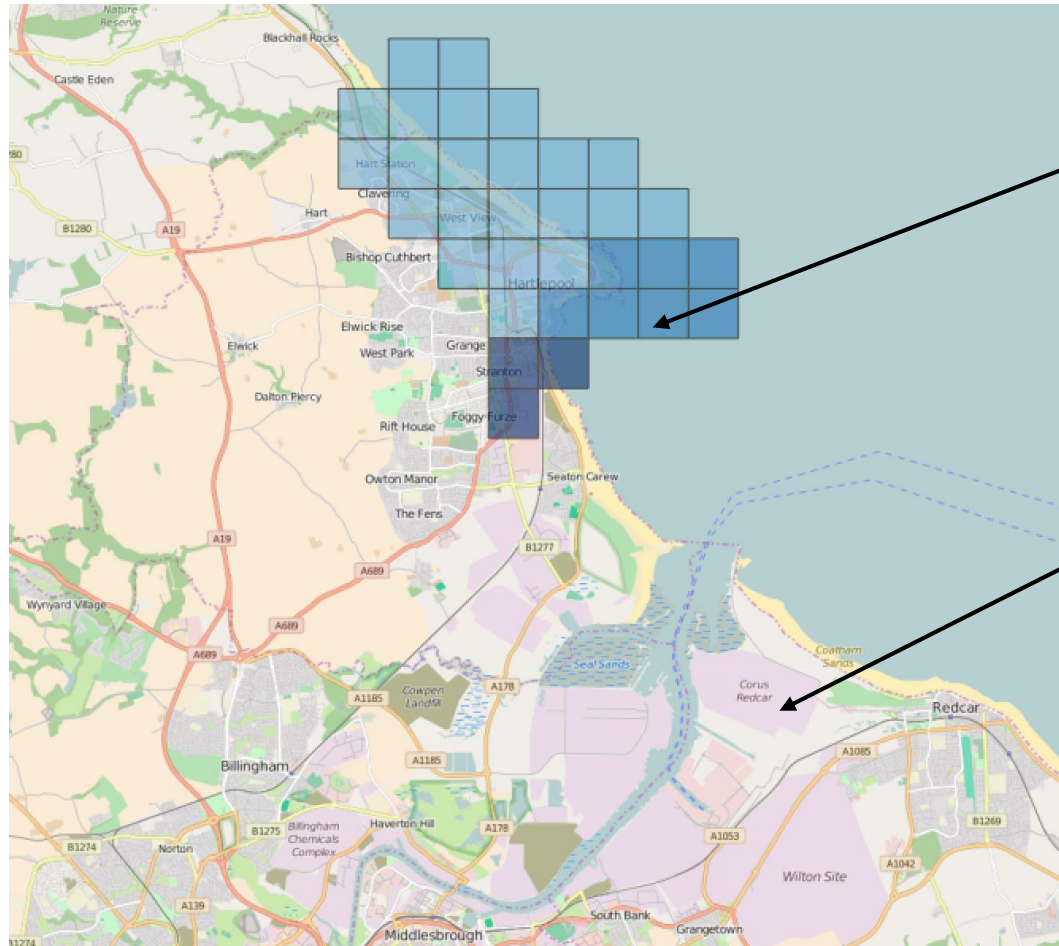
Confidence



40 - 80%

80 - 95%

95%+



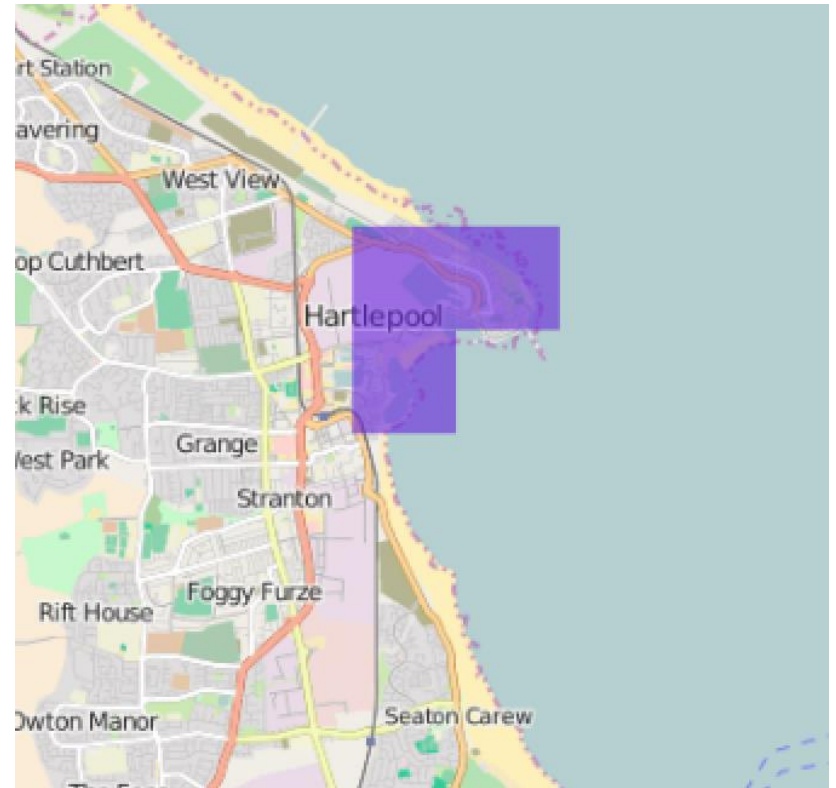
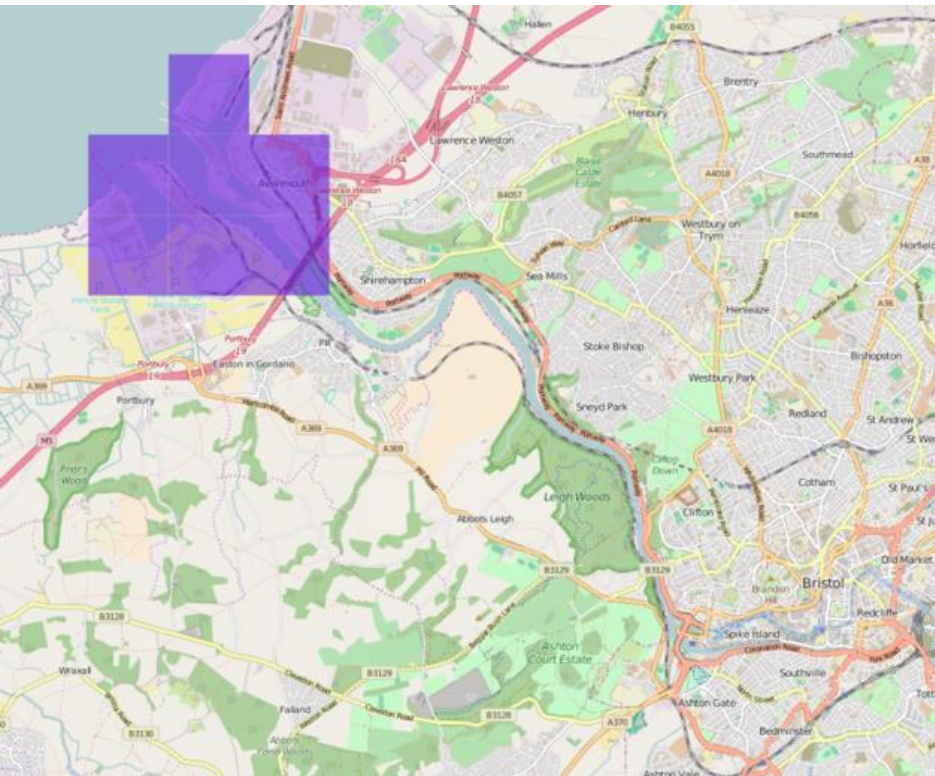
Suggests
this is a
port

Suggests
this is not
a port

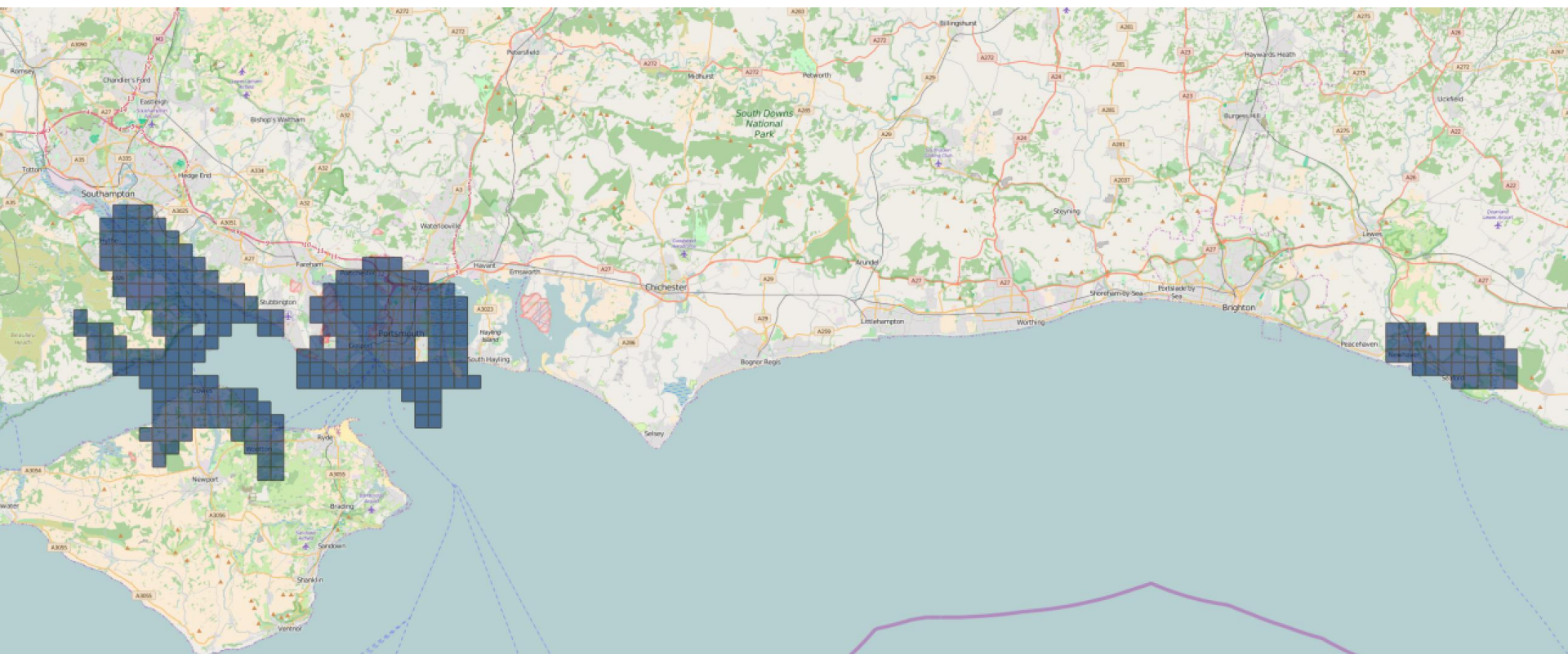


Additional training?

- Bristol and Hartlepool



Artificial Neural Network: outputs



Conclusion

- Simple distance measures produce some meaningful results
- Machine learning approaches show great promise
- Refinement in the training and classification of ports required
- Further evaluation required
- Research ongoing (as they say)



References

Alderton P 2008 Port Management and Operations. *Informa*.

Hofmann, M & Klinkenberg, R 2013 “RapidMiner: Data Mining Use Cases and Business Analytics Applications (Chapman & Hall/CRC Data Mining and Knowledge Discovery Series),” *CRC Press*

Noy, NF & McGuinness, DL 2016 Ontology Development 101: A Guide to Creating Your First Ontology

