

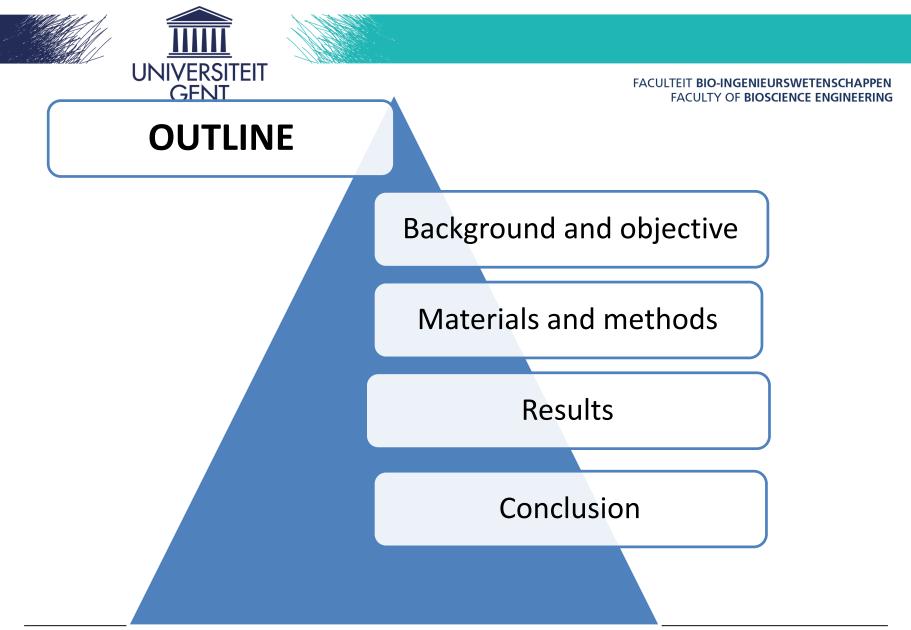
# Macroinvertebrate based mathematical models for the prediction of microbial pathogen in rivers

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BIG DATA AND DATA MINING 2016, London, England, Date (26/09/2016)

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#### 1-BACKGROUND AND OBJECTIVE (I)

- The quality of the water must accomplish standards.
  - drinking water, recreational purpose, Irrigation
- The indicators used to verify microbial contamination of water are:
  - total coliforms and fecal coliforms and/or Escherichia coli



#### 1-BACKGROUND AND OBJECTIVE (II)

- Checking the fulfillment requires expensive and highly trained personnel in laboratories
- Biota works as a permanent monitor of water
- biological samples can :
  - reflect an increase in pollution.
  - predict average values of chemical parameters



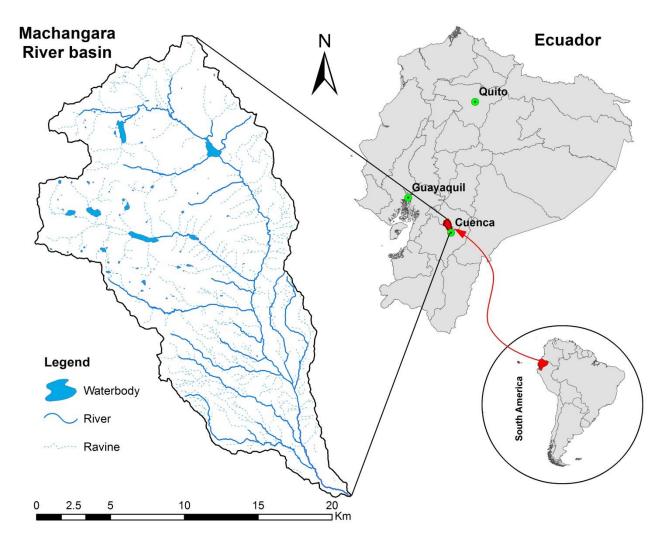
# Objective

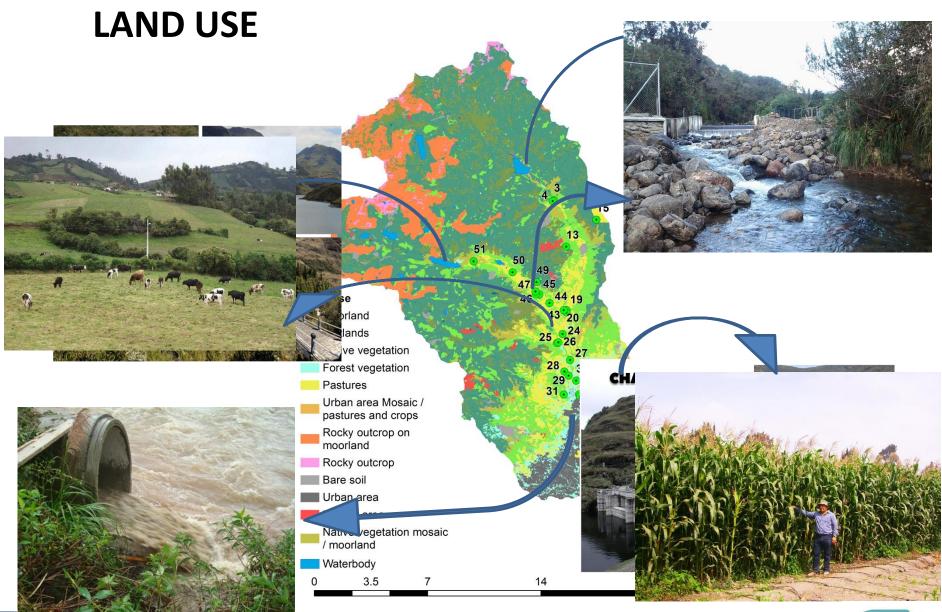
 Introduce a quick way of checking the fulfillment of fecal coliforms standards using macrobenthos.

 Analyzes the requirement to include biology and hydro-morphology aspects in Ecuadorian regulations to assess river ecosystem health.

#### 2- MATERIALS AND METHODS

#### LOCATION









# DATA COLLECTION:

- Completed information on 33 locations

# Physicochemical, hydraulic, microbiological

- Laboratory
  - BOD<sub>5</sub>, COD, Nitrate + Nitrite, Ammonia, Organic Nitrogen, Phosphates, Total Phosphorus, Fecal and Total Coliforms, Real Color, Turbidity, Total Solids
- **Field:** Flow Velocity, Ph, Conductivity, Temperature, Dissolved Oxygen

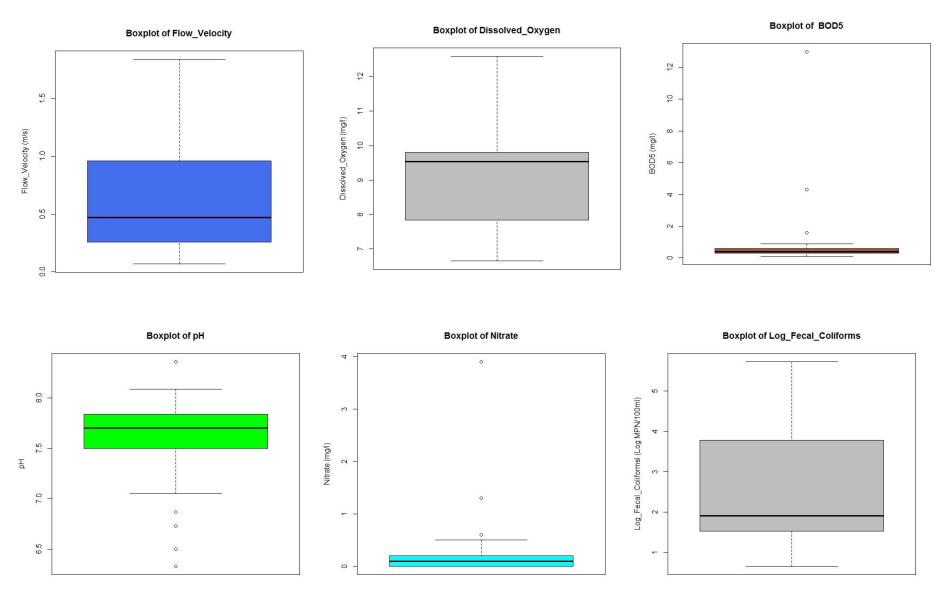
# Macrobenthos

39 families (taxa) found





#### **Variables Variation**





 Ecuadorian Water Quality Regulation for Fecal Coliforms

Regulations	Water used to	Fecal Coliforms Limited Value MPN/100 ml		
First	Recreational with primary contact	≤200		
Second	Agriculture and Livestock	≤1,000		
Third	raw water previous to non-conventional treatment*	≤2,000		
* Conventional treatment refers to chemical addition, rapid mixing, flocculation and sedimentation				

# ECOLOGICAL WATER QUALITY

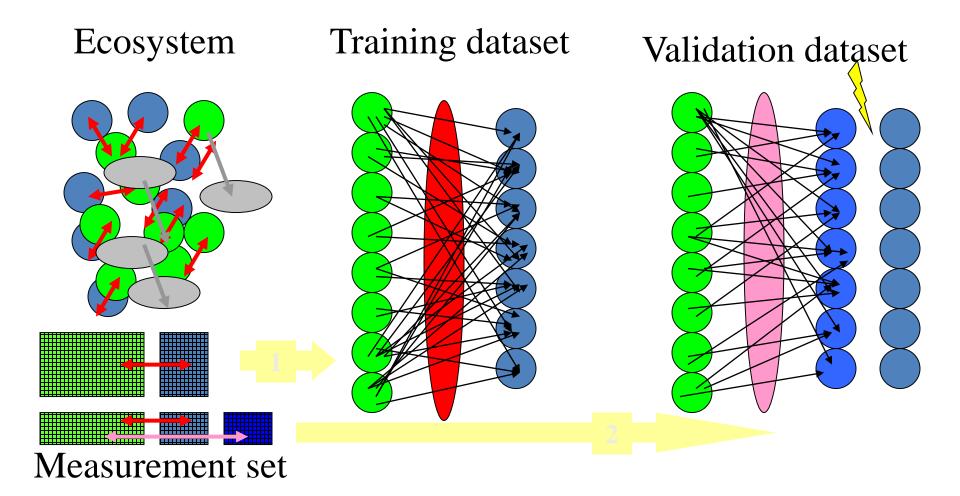
- Biological Monitoring Working Party Index Col
- BMWP-Col = f(Sensitivity of Macrobenthos)
- Sensitivity -> 1-10 (Low High Sensitivity)

Class	Quality	BMWP	Color
1	Very Good	> 100	
Ш	Good	61 - 100	
Ш	Moderate	36 - 60	
IV	Deficient	16 - 35	
V	Bad	≤ 15	

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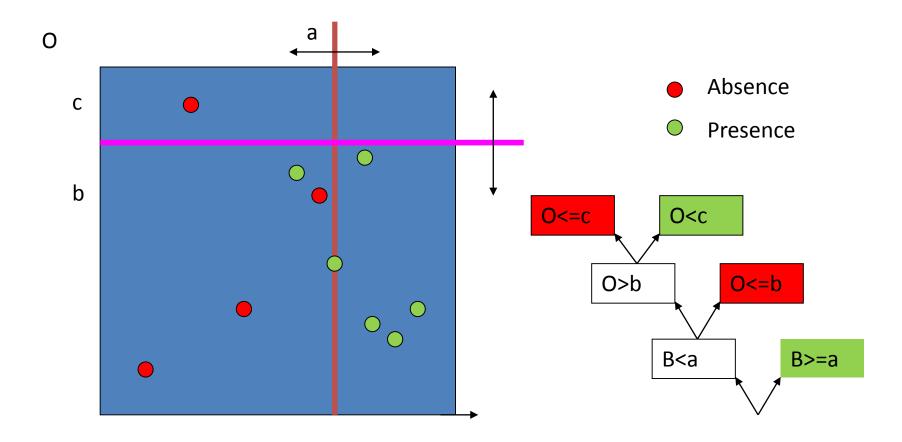
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#### **Model development**



### Methodology Pruned Multi-target Clustering Trees (PMCT)

 Classification trees: searching for if-then rules (threshold values): 100% reliable and 'safe' models





#### **Model Performance**

- Models must be evaluated based on statistical and ecological criteria.
- Models must be as clear and simple as possible.

#### Settings

- Machine learning software: Waikato Environment for Knowledge Analysis (WEKA)
- Three, five, ten-fold cross validation (k fcv)
- Pruning process
  - Pruning confidence factors (PCF): 0.25, 0.10



#### **Model Performance**

- Confusion matrix from Decision Tree Models:

		Predicted Class	
		Yes	No
Observed	Yes	TP	FN
Class	No	FP	TN

- Correctly Classified Instances (CCI):> 0.70
- Cohen's Kappa Statistic: > 0.40
- Lowest value for the false negative (FN) will increase sensitivity



- Cost sensitive classifier (CSC): gives new weights in training instances FN, FP
- Overall confusion entropy of a confusion matrix (CEN): evaluates the confusion level of the class distribution of misclassified samples:

$$CEN = (P_1 + P_2)CEN_j$$

 $P_1 = \frac{TP + FN}{2(TP + FN + FP + TN)} and P_2 = \frac{FP + TN}{2(TP + FN + FP + TN)}$ 

$$CEN_j = -P_{FN}log_2P_{FN} - P_{FP}log_2P_{FP}$$



# Model Optimization:

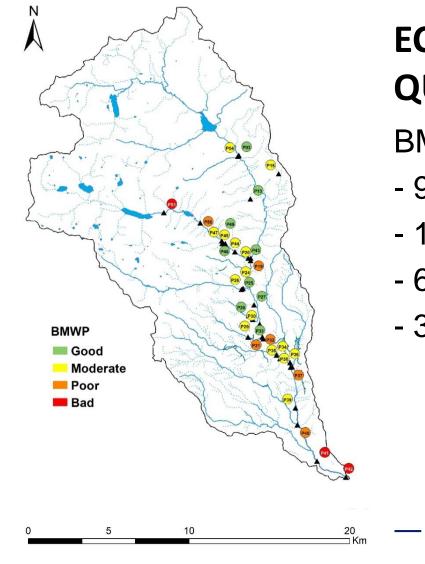
- $-P_j$ : confusion probability of class j
- $CEN_j$ : confusion entropy of class *j*.

$$P_{FP} = \frac{FP}{FN + FP + 2TP} \text{ and } P_{FN} = \frac{FN}{FN + FP + 2TN}$$

- $P_{FP}$  and  $P_{FN}$  are the misclassification probability of classifying the samples of class *i* to class *j* subject to class *j*
- Higher accuracy corresponds to lower confusion entropy

# 3- **RESULTS**

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# ECOLOGICAL WATER QUALITY

**BMWP-Col**:

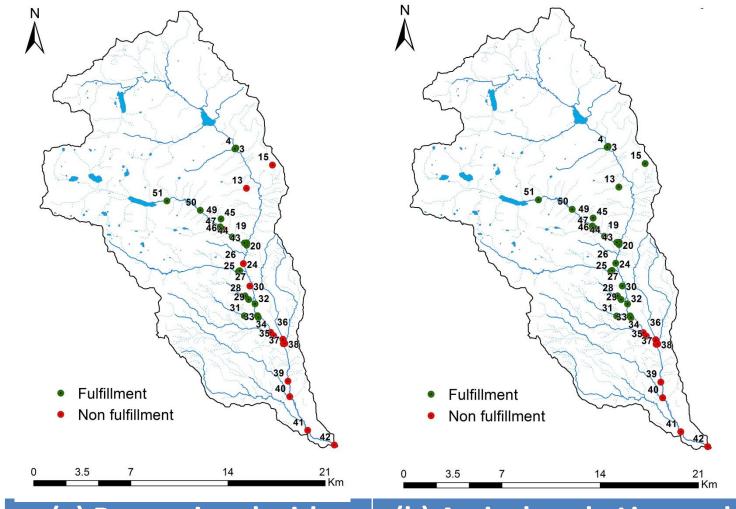
- 9 good
- 15 moderate
- 6 poor
- 3 bad



#### Analysis of Ecuadorian Water Quality Regulation for Fecal Coliforms in relation to BMWP-Col

BMWP-Col	First Preservation Regulation	Second Preservation Regulation	Third Preservation Regulation
Bad	2	2	2
Deficient	2	2	2
Moderate	8	5	5
Good or very good	2	-	-
TOTAL non fulfillment	14	9	9

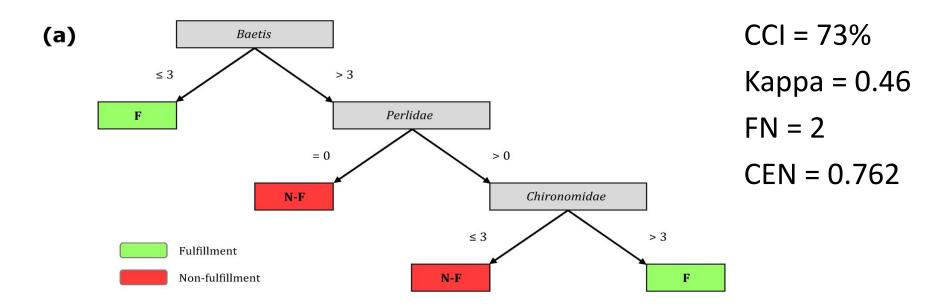
# Fulfillment of Fecal Coliforms limits in relation to water use



# (a) Recreational with primary contact

(b) Agricultural - Livestock use and (c) raw water

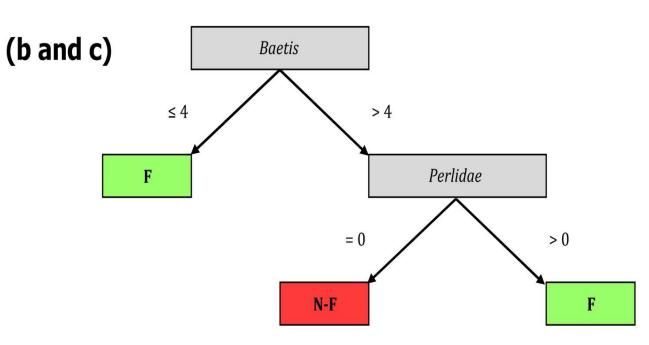
# Model Performance (1) First Model: Primary Contact – Fecal Coliforms Regulation.



*Chironomidae* families include species with large differences in tolerance to pollutants

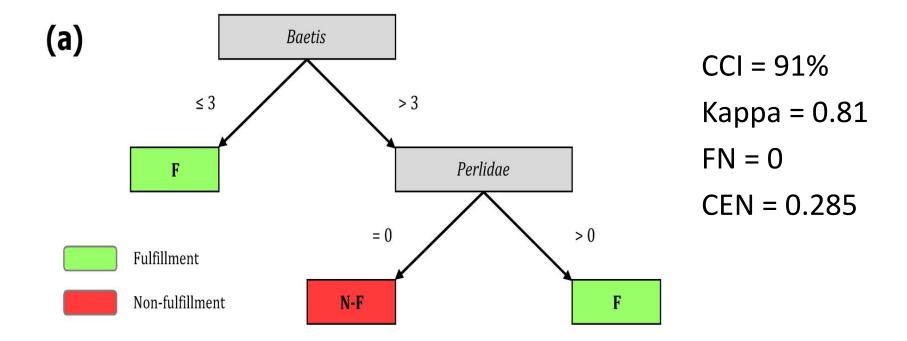
# Model Performance (2)

**Second model:** Agriculture - Livestock Irrigation and raw water previous to nonconventional treatment



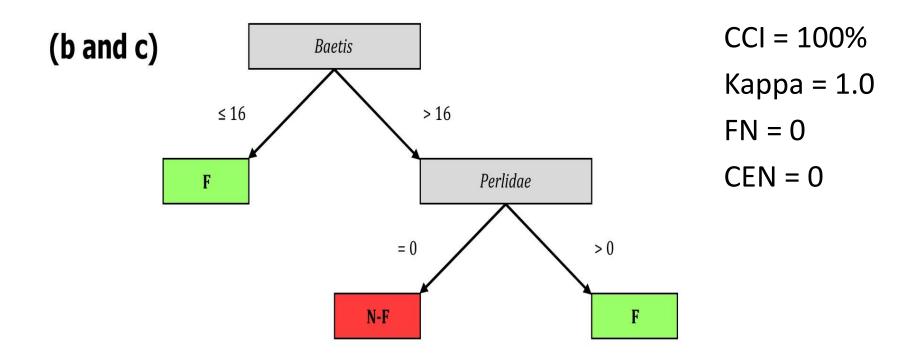
CCI = 94% Kappa = 0.85 FN = 1 CEN = 0.348

#### Model Optimization (2) First Model: Primary Contact – Fecal Coliforms Regulation.



# Model Optimization (3)

**Second model:** Agriculture - Livestock Irrigation and raw water previous to nonconventional treatment



- Four models were selected.
  - Two Models: fecal coliforms threshold in recreational with primary contact water use.
  - Two Models: fecal coliforms limits in agricultural
    livestock water use, or raw water for drinking water treated with non-conventional processes.
- The cost-sensitive classifier (CSC) in the Weka can reduce false positives (FP) in the confusion matrix, improved the reliability of the resulting models.
- Confusion entropy of a confusion matrix (CEN) was lower when the confusion matrix had lower FN values.



#### Acknowledgement

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# Thank you



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