

Nowcasting of fecal coliforms presence using an artificial neuron network in India

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Research motivation - Global safe water shortage

- **2 billion**
Lack safely managed drinking water
- **771 million**
Without basic drinking water services
- **159 million**
Drink water directly from surface water
(streams or lakes)
- **800 children**
Under 5 years of age die daily because
of using unsafe drinking water



Research motivation - Global safe water shortage

Population expansion,
urbanization and industrialization

Increased effluent
production

Insufficient sewage
infrastructure and
water treatment plants

Contaminated
water sources





Drinking water quality monitoring

- Water quality should be routinely analyzed
- Many kinds of parameters can be monitored in real-time
- Accurate information of biological contamination is an integral part of water systems management

Fecal coliforms monitoring

- Fecal coliforms (FC) or Escherichia coliform should be tested
- WHO standard is 0 MPN/100ml
- Traditional monitoring test takes at least 18-24h
- In rural communities the test routine is infrequent – up to 6 months
- Real-time monitoring is expensive



The Goal

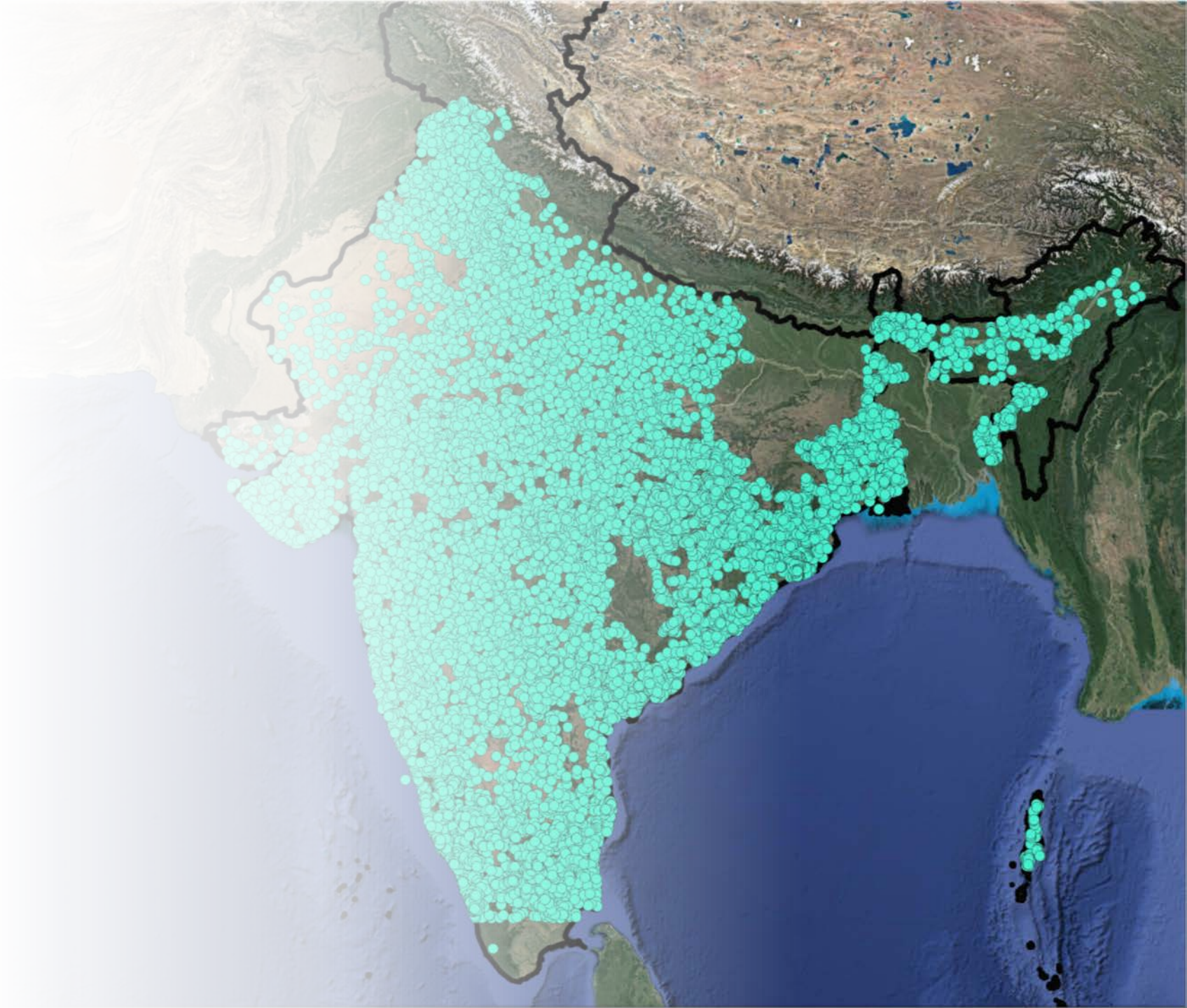
- Find a cheap, real-time and in-situ solution for FC monitoring





The Data

- Was collected by the Central Pollution Control Board of India
- From throughout India
- Various kinds of sources
- A large number of parameters



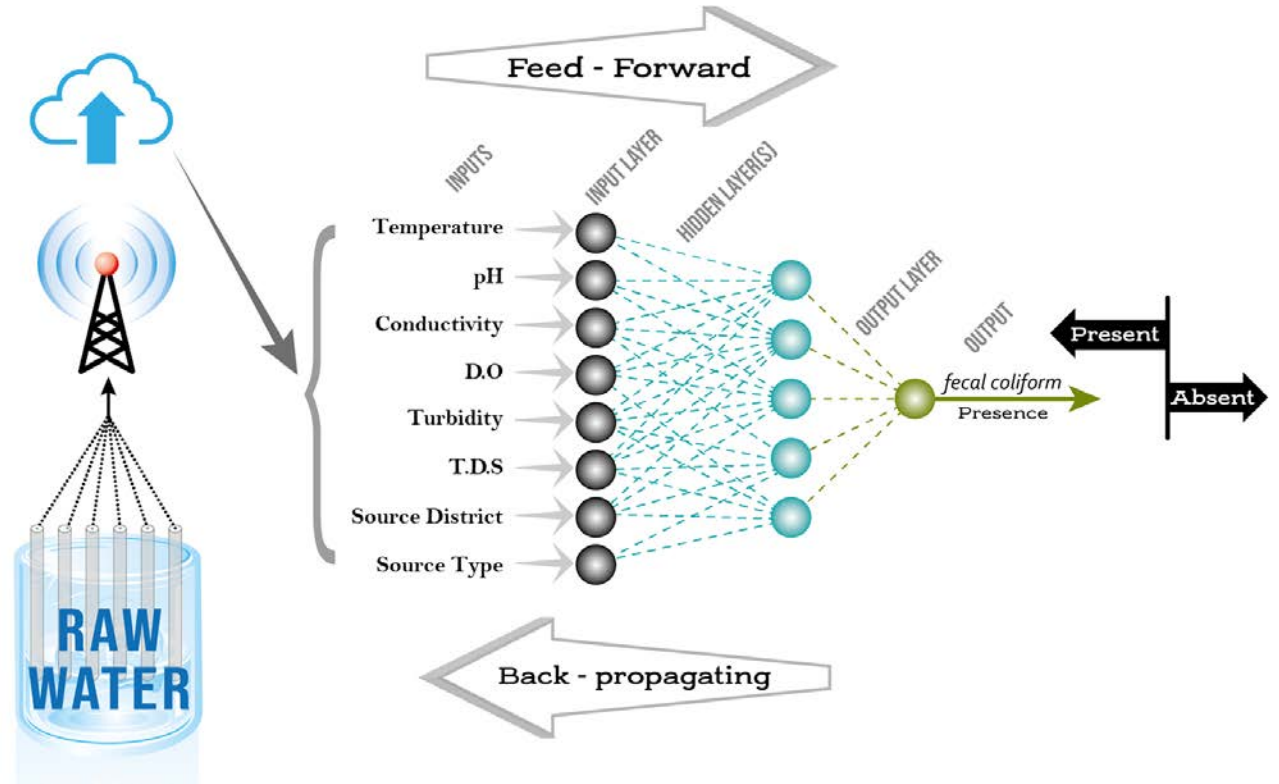
Our solution

- FC sensor based on machine learning or artificial intelligence, which uses easy-to-monitor, real-time and *in situ* parameters as inputs
- Gives a ‘nowcast’ of FC absence or presence in water sources
- Can be later upgraded for more parameters



Method

- Using a multilayer perceptron model
- The model is fed in with the following low-cost and easy to monitor parameters:
 - Temperature
 - pH
 - Electrical conductivity
 - Total dissolved solids
 - Dissolved oxygen
 - Turbidity
 - Geographical parameters (Type of source, and location)







Results – Confusion matrix

$$\text{Accuracy} = \text{TP} + \text{TN} = 92\%$$

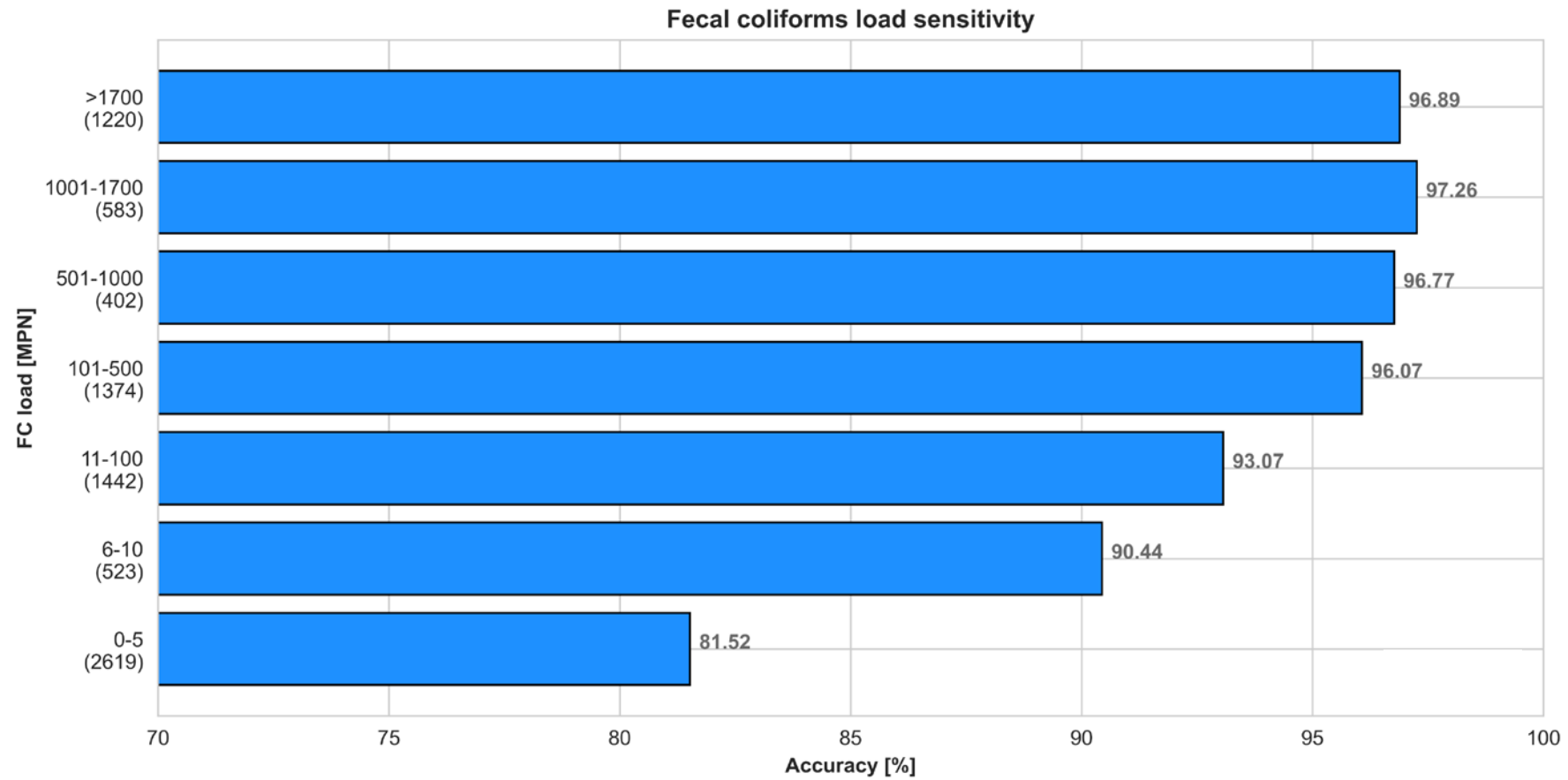
$$\text{Sensitivity} = \frac{\text{TP}}{\text{FN} + \text{TP}} = 98\%$$

$$\text{Precision} = \frac{\text{TP}}{\text{FP} + \text{TP}} = 93\%$$

$$\text{Specificity} = \frac{\text{TN}}{\text{FP} + \text{TN}} = 75\%$$

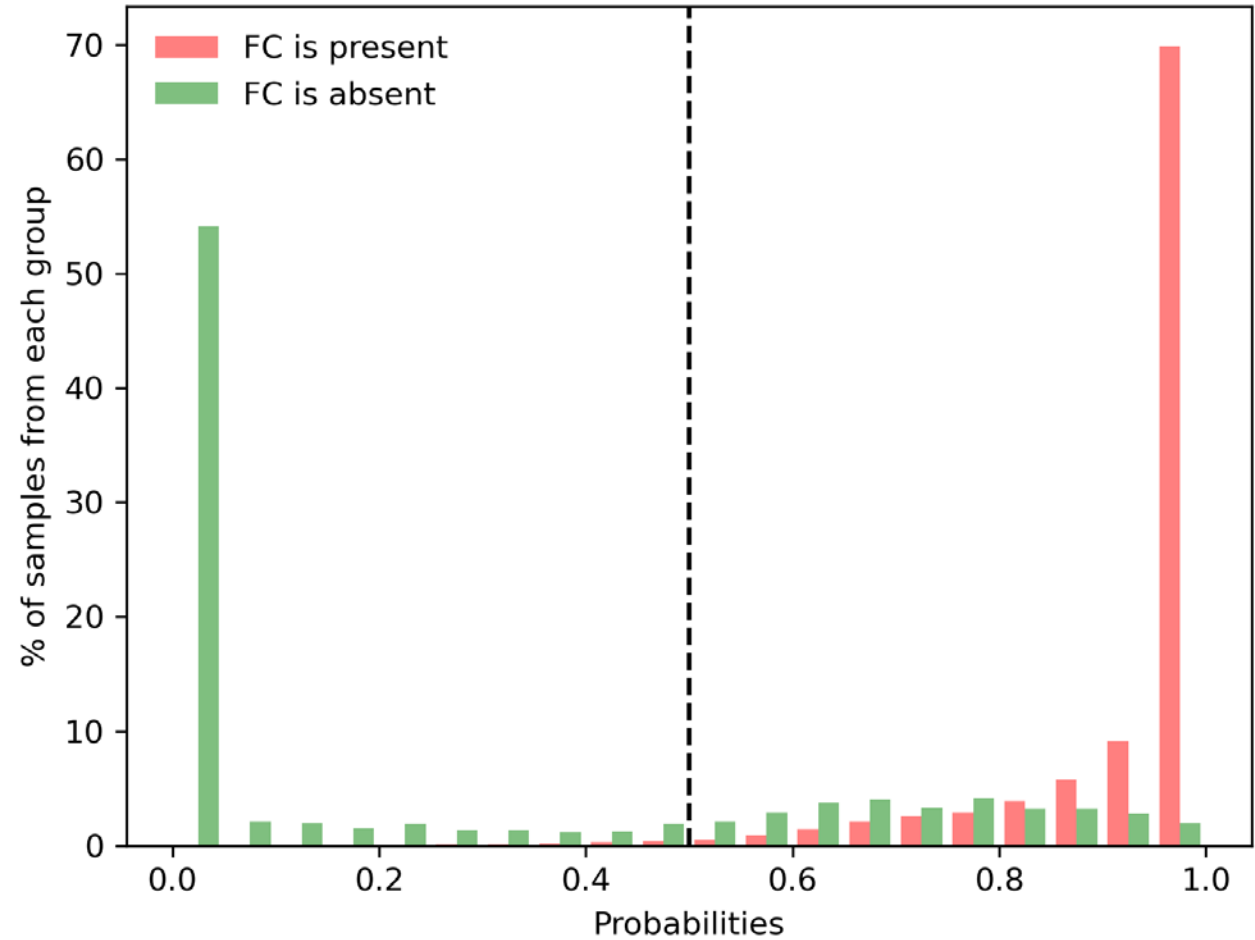
True label	Absent	TN No problem 15.25% 	FP False alarm 5.19% 
	Present	FN Worst case 2.08% 	TP Prevention 77.48% 
		Absent	Present
		Predicted label	

Results – Biological load effect on sensitivity







Results – Probability histogram

- The probability of a sample to be present







Results – Sensitivity–specificity–precision tradeoffs

- when raising the cutoff to 0.75:

- Accuracy: 90.56% 
- Sensitivity: 92.66% 
- Precision: 95.34% 
- Specificity: 82.36% 

- when lowering the cutoff to 0.25:

- Accuracy: 92.54% 
- Sensitivity: 99.74% 
- Precision: 91.63% 
- Specificity: 64.53% 

True label	Absent	TN 13.2% (1017)	FP 7.3% (559)
	Present	FN 0.02% (16)	TP 79.4% (6118)
		Absent	Present
		Predicted label	

Summary

- We showed the feasibility of nowcasting FC contamination:
 - About 93% of accuracy
 - 98% of sensitivity
 - 93 of precision
- By moving the threshold, we can “trade” sensitivity with precision



Thanks for listening
Drink water and keep safe

