

Groundwater Monitoring Using Handpump Data in Rural Africa

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Overview

- | Monitor groundwater depth using handpump vibration data.
- | Builds on [1, 2, 3, 4] to track handpump usage and facilitate handpump maintenance systems in rural Kenya.
- | Cost-effective and scalable infrastructure where traditional technology/data may be limited or non-existent.
- | Intended to complement existing hydrogeological modelling.

Method

- | Combines novelty detection with regression approach (LSTM).
- | Designed alongside a handpump maintenance infrastructure.
- | Can also incorporate hydroclimatic and hydrogeological data.

Results

- | The novelty scores of training (blue dots) and test (red dots) examples are represented by the log-likelihood of examples given the normal model.
- | Usually the novelty scores appear to be higher (i.e. the log-likelihood values are lower) immediately following a repair (dashed black lines).
- | The post-repair vibration data are calibrated to match the pre-repair data.
- | The estimates for training, validation, and test sets are shown in blue, orange, and red respectively.
- | A summary of results (estimation errors in metres) for all handpumps for both MLP and LSTM techniques are provided in the Table.
- | The fractional change in water columns at the boreholes at two locations are shown with respect to a common reference date.

| | SP1 | SP2 | MP1 | MP2 | MP3 | DP1 | DP2 |
|------|------|------|------|------|------|------|------|
| MLP | 0.14 | 0.26 | 0.31 | 0.29 | 0.18 | 0.60 | 0.85 |
| | 0.12 | 0.19 | 0.15 | 0.11 | 0.10 | 0.34 | 0.71 |
| LSTM | 0.09 | 0.12 | 0.23 | 0.14 | 0.15 | 0.43 | 1.07 |
| | 0.04 | 0.06 | 0.17 | 0.11 | 0.12 | 0.47 | 0.41 |

Discussion

- | Frequent handpump breakdown and subsequent repair complicates learning a consistent model.
- | Current solution uses novelty scores and vibration data calibration.
- | In future, explore more principled methods, e.g. transfer learning.
- | Designed to be implemented at scale using a network of pumps.
- | In future, explore multi-task LSTM to model multiple handpumps simultaneously and fuse hydro-climatic and hydrogeological data.

Ethical Considerations

- | Although intended to enable sustainable groundwater management among competing users (e.g. community vs. industry), incompetent management poses risks to vulnerable population.
- | The data may also unintentionally induce forced migration of households out of areas abundant in groundwater resource.
- | A successful implementation relies on both adequately training local experts as well as ensuring sound groundwater governance.

References

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- [4] H. Greeff, A. Manandhar, P. Thomson, R. Hope, and D. A. Clifton. Distributed inference condition monitoring system for rural infrastructure in the developing world. *IEEE Sensors Journal*, 19(5):1820–1828, March 2019.