Predicting Fertilizer Input for Rice Cultivation in India

Home to over 1.38 billion people, India is tackling a severe hunger crisis. Though the country has achieved self-sufficiency in grain production, nearly 14% of the population is still undernourished. India's agricultural landscape is primarily rural, where widespread poverty, low literacy rates, and poor infrastructure lead to questions over its sustainability. **Indiscriminate use of fertilizers has led to significant irregularity in crop production despite consistent agricultural subsidies.**

With the current global shortage of fertilizers, precision farming is vital to eliminate redundant costs and streamline resources to ensure equitable food access for all communities. Here, we assist policy-makers in their decisions through models predicting the fertilizer consumption (nitrogen, phosphorus, and potash) required to obtain a specific rice yield.

Rice is the staple food crop for over 50% of India's population. Being a hardy crop, rice can be cultivated on a variety of soils, including clays, silts, and gravel. Typically, rice crops are planted at the onset of the monsoon season (June – July) and harvested during November – December every year.

The International Crops Research Institute for the Semi-Arid Tropics (ICRISAT) maintains a <u>district-level database</u> that contains up to 26 years (1990 – 2016) of India's rice cultivation and environmental data. We segmented India into six different clusters for rice cultivation by utilizing



seasonal values of various ecological features (precipitation, runoff, wind speed, minimum and maximum temperature). All districts within the same cluster share similar environmental characteristics, thereby permitting a joint analysis of their historical rice yield data.

Our data, while quite noisy, reveals a fluctuating relationship between the historical mean fertilizer input per unit area and the resulting rice yield. Consequently, for a given target rice yield based on population requirements, our models enable decision-makers to place intelligent cluster-specific budget constraints on fertilizers, and minimize their indiscriminate use.

Future extensions to our model will incorporate soil nutrient data, solar irradiance, and knowledge of off-season farming practices (e.g., crop rotation) to improve the accuracy of our estimated fertilizer inputs.