

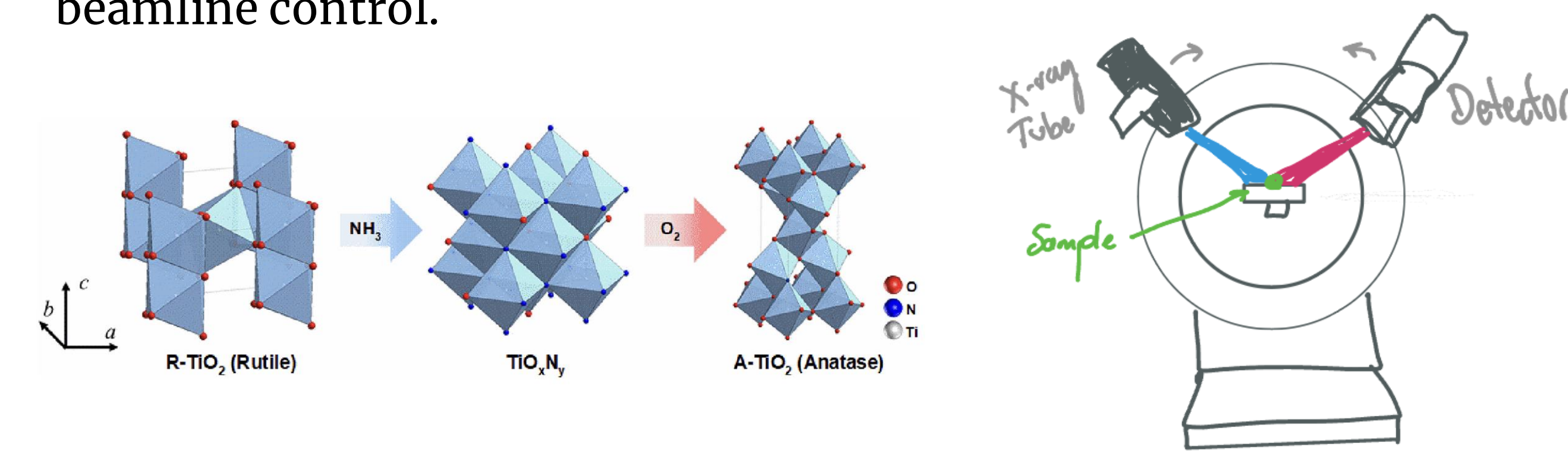
# AUTONOMOUS EXPERIMENT DESIGN: Machine-Learning Guided X-Ray Powder Diffraction for Efficient Phase Tracking

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## INTRODUCTION

- Motivation:** Synchrotron experiments demand time, resources, and constant operator input.
- Goal:** Develop a Python-automated, ML-driven system for autonomous beamline control.

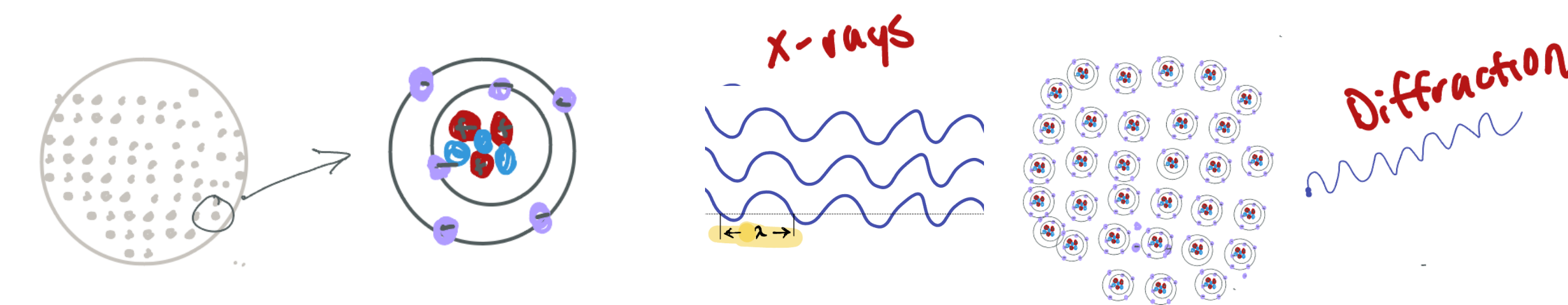


### Research Questions

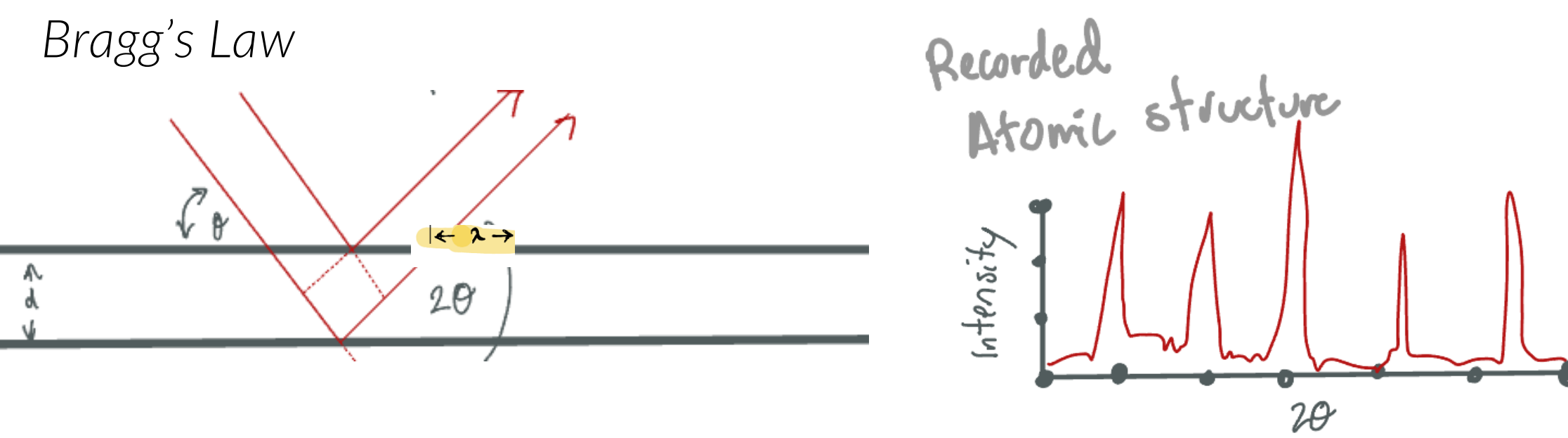
- How can we direct the computer to scan only information-rich regions?
- Can intelligent scan selection save time, storage, and materials *without losing structural data*?

## BACKGROUND

**X-rays** have wavelengths comparable to atomic spacings, making diffraction a powerful probe of crystal structure.



**Powder diffraction:** X-rays scatter from tiny grains; a two-detector setup records unique diffraction patterns.

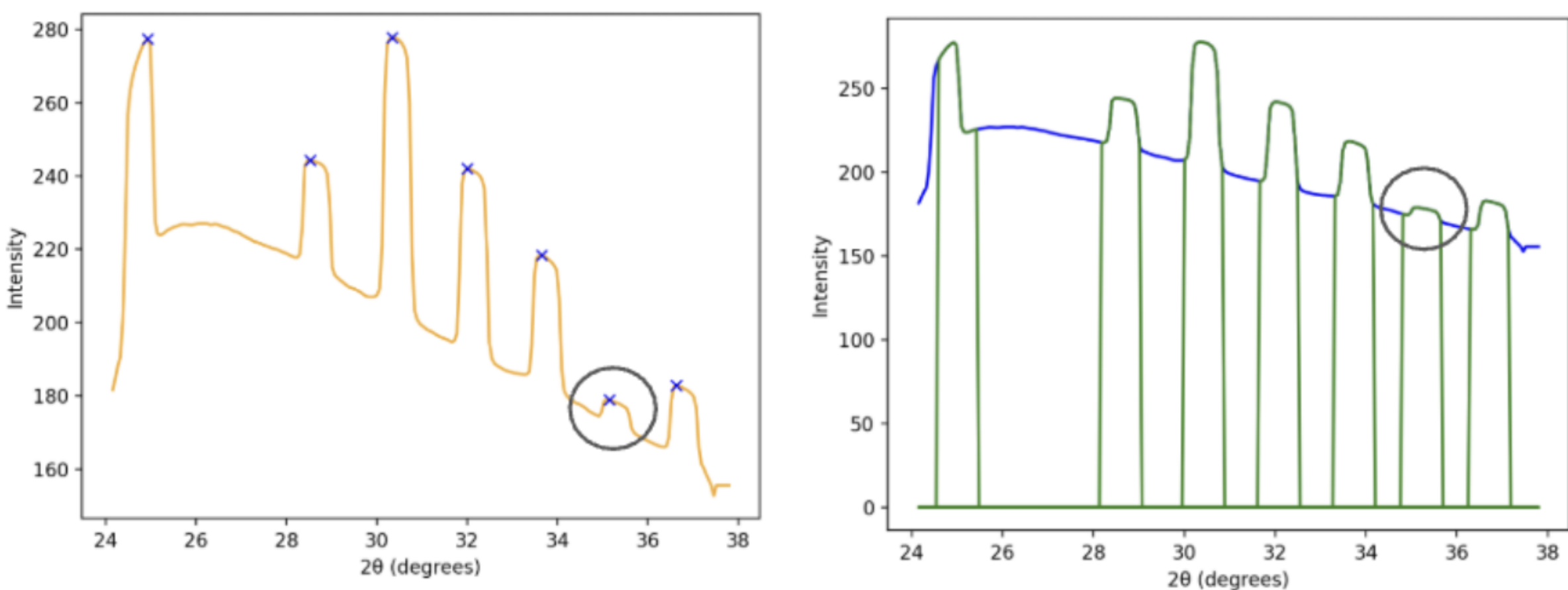


**References:** [1] Cullity & Stock, *Elements of X-Ray Diffraction*, 3rd ed., 2014. [2] Diebold, *Surf. Sci. Rep.*, 48, 53–229 (2003). [3] Sivia & Skilling, *Data Analysis: A Bayesian Tutorial*, 2nd ed., 2012. [4] SSRL BL 2-1 User Documentation, 2025. [5] Pease, D. M., “What is X-Ray Diffraction?” YouTube, 2020.

## METHODS

### Workflow Development

- Six scanning strategies (top peaks, intensity change, shifting, growth, shrinkage, max-change regions).

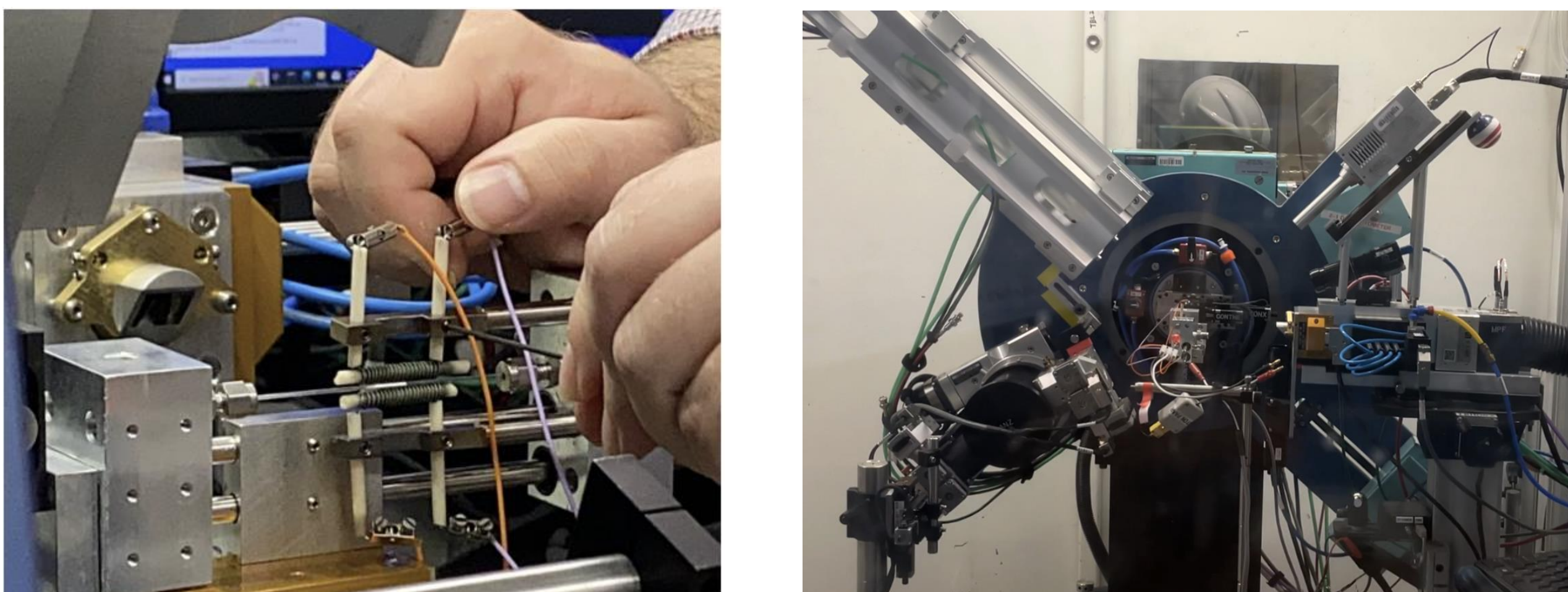


- Integrated low-res → 1D patterns; local maxima & peak thresholds.
- High-res scans only where meaningful change is expected.



### Beamline Setup (SSRL 2-1)

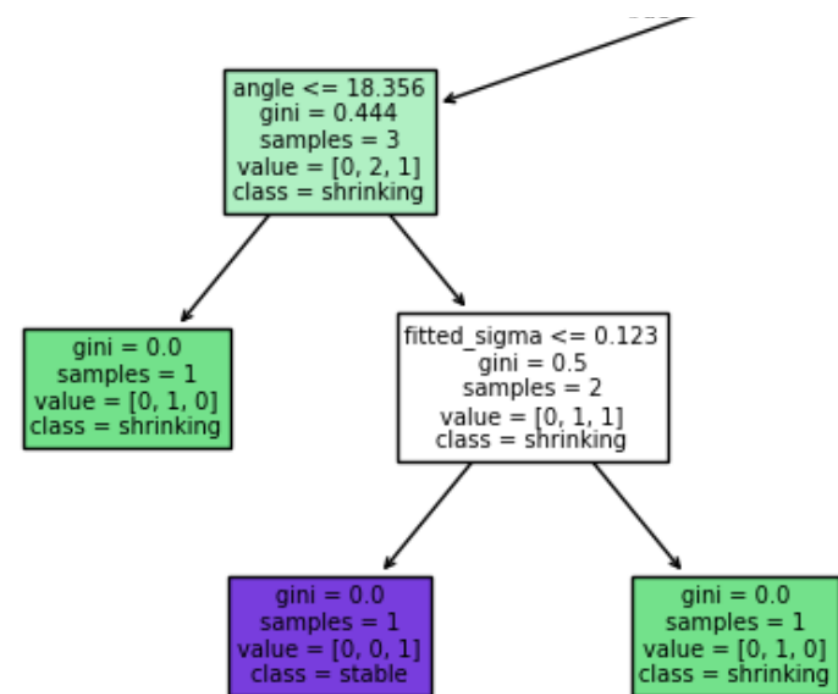
- TiO<sub>2</sub> nano powder in quartz capillaries.
- Controlled heating (20 °C → 820 °C).
- Dual-detector system: fast continuous monitoring + high-resolution targeted scans.



## RESULTS

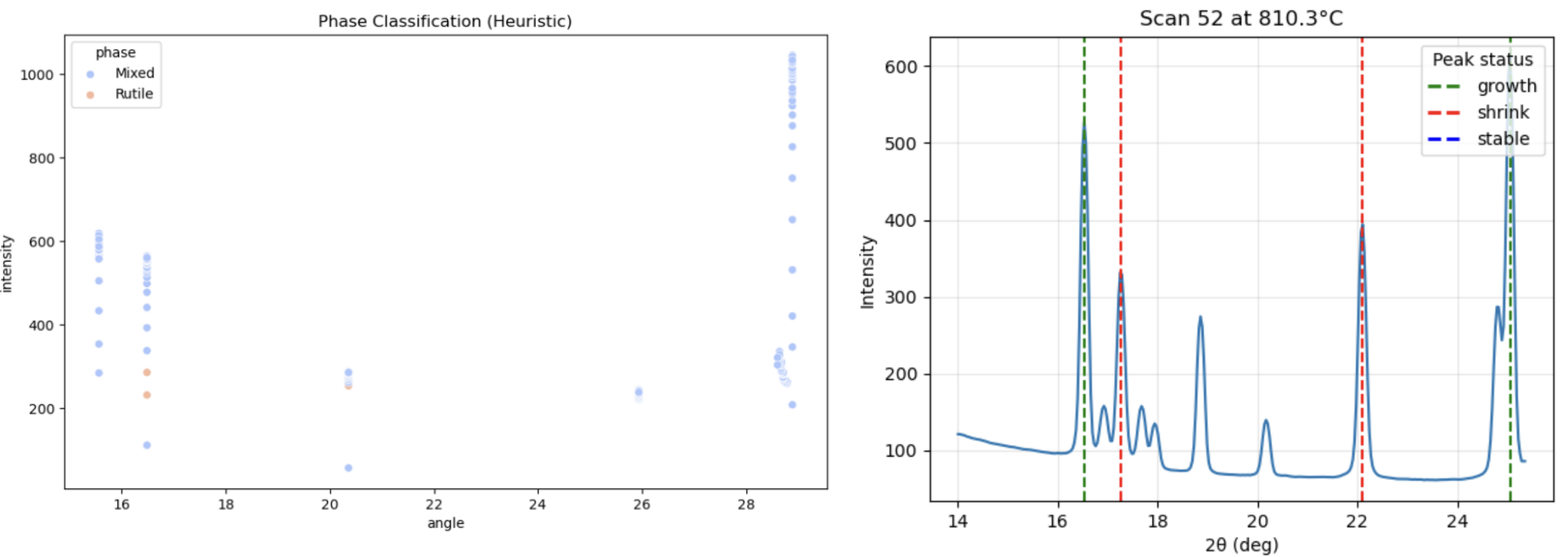
### Algorithm Improvements

- Peak classification: growing, shrinking, stable.
- Phase assignment (anatase vs rutile).
- Gaussian fits for precision.



### Machine Learning Integration

- Decision Tree Classifier** trained on labeled peaks.
- Prioritized scans of relevant phase changes.



- Beamtime efficiency ↑:** Reduced unnecessary scans.
- Phase tracking:** Captured anatase → rutile transition.
- High-quality data:** Combined raw intensity + categorical descriptors.
- Adaptability:** Framework extendable to other materials.

## CONCLUSION

- ML-driven scan selection improves efficiency and scientific value of synchrotron experiments.
- Interpretable models (decision trees) enhance reliability and real-time decision-making.

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