

Ubiquitous DAS

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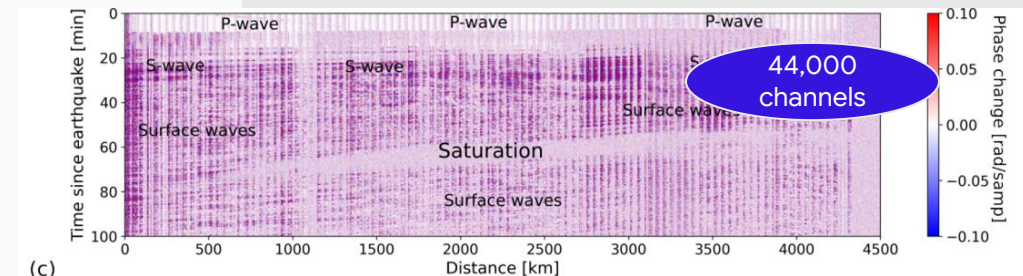
What if DAS coverage were
ubiquitous?

What processing components need to
be in place to support it?



Ingredients for ubiquity are being worked on

- Price of interrogators is (will be) more affordable with use of less expensive components and volumes. Especially if fit for purpose trade offs are applied.
- Compute price is dropping and hardware footprint is reducing.
- Interrogator specs, e.g. range, noise floor, .. are improving.

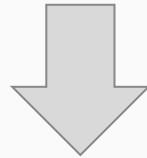


M. Mazur et al., "Submarine Cable Deep-Ocean Observation of Mega-Thrust Earthquake and Tsunami with 44,000 100-m Spaced Sensors," ArXiv e-prints, Art. No. 2509.24813, Sep. 2025.

Ubiquity leads to massive volumes of data generated

In the US, 9.3M miles of fiber to are expected to be available by 2028*

Assume: 1000 Hz acquisition, 5m spacing, 2 byte precision

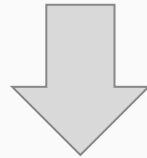


517,000 TB / day

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Solutions that need to be built

- 1. Machine learning
- 1. Edge processing
- 1. Compression

Drowning in data. Thirsty for insights. Solution: ML

Noisy data problem: **Terabytes / day** from a single cable, but mostly “noise”.

Requirements:

- Balance sensitivity and specificity - lots of false triggers make manual reviews infeasible
- Quickly adaptable to new sites and new noise regimes
- Real-time or near real-time detection

Operational constraints:

- Equipment cost
- Data storage cost
- Bandwidth limitations, especially for remote sites
- Space and power budget limitations on a platform



Gemini

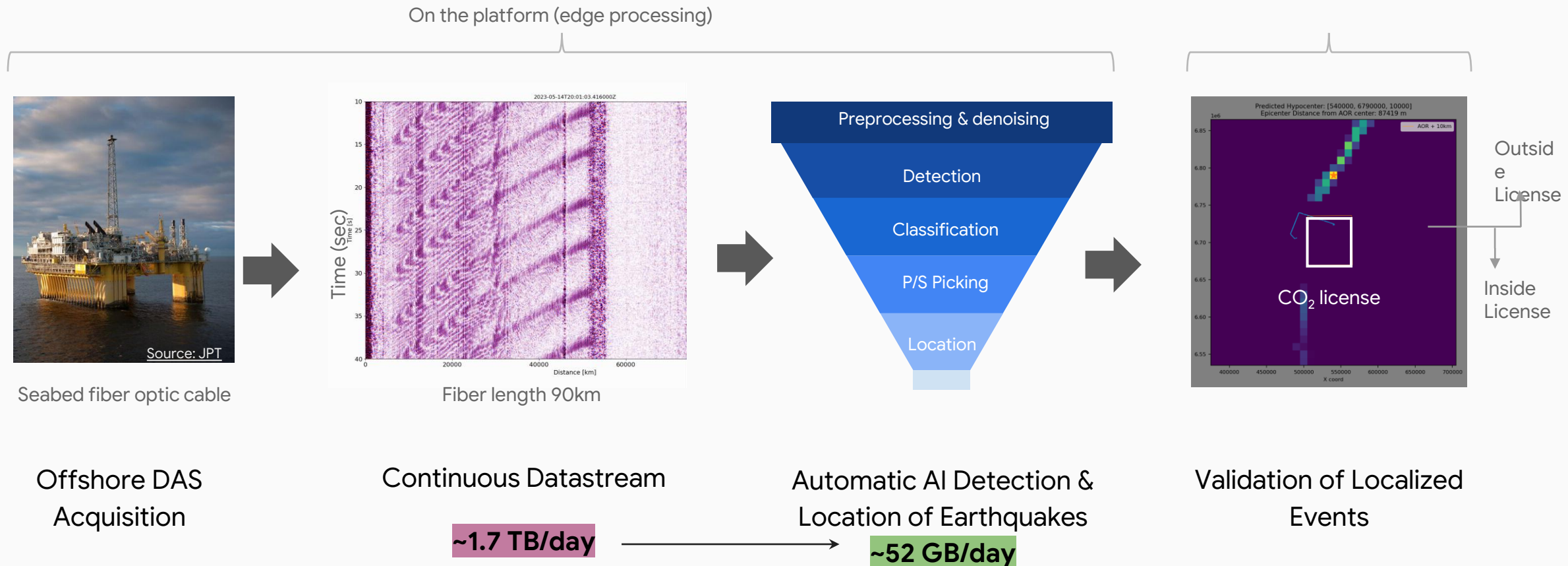
3%

In a typical site, less than
3% of data contains a
useful signal.

Real-time edge microseismic monitoring system

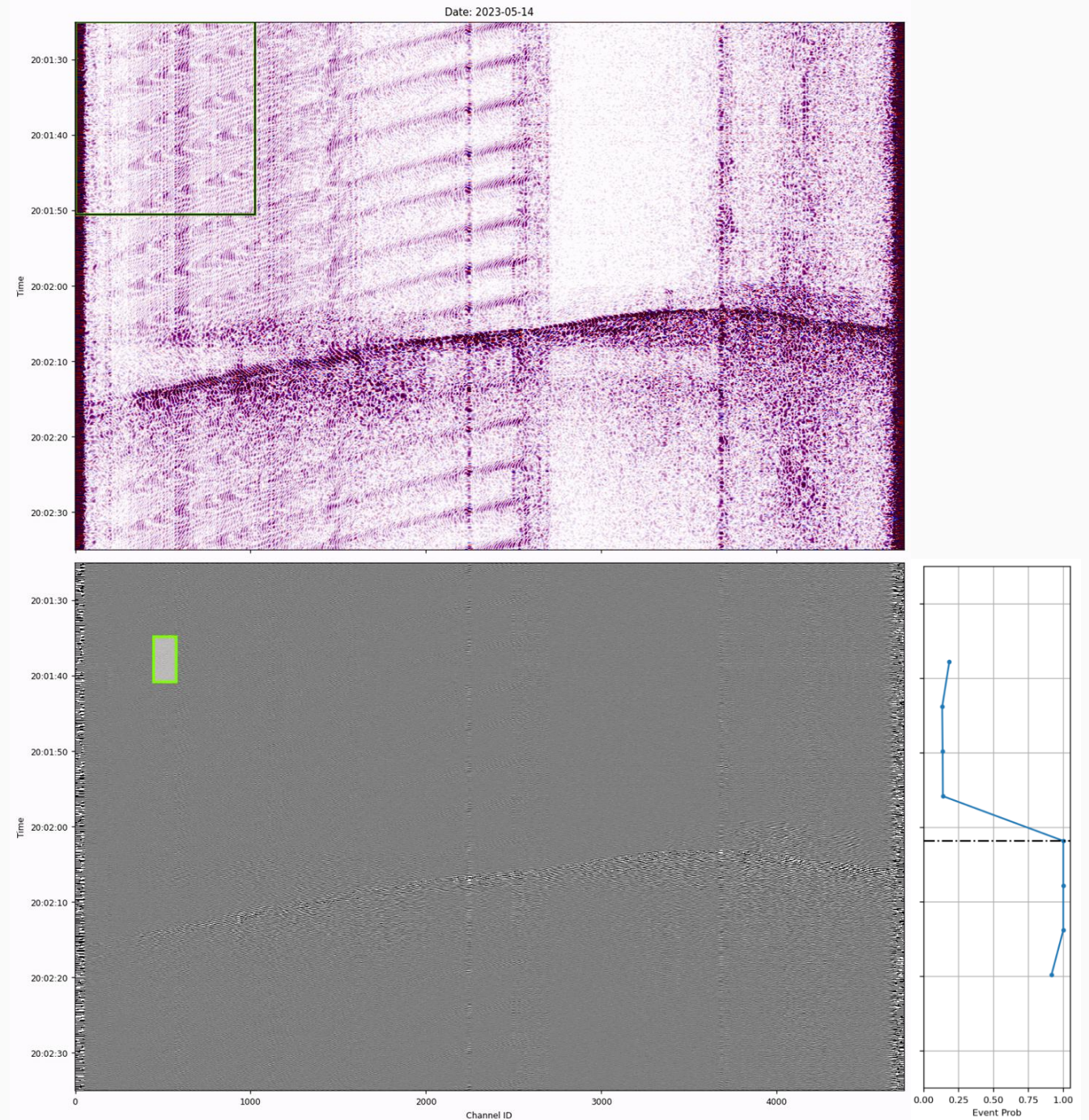


An operational solution has been developed to manage the huge DAS data flow

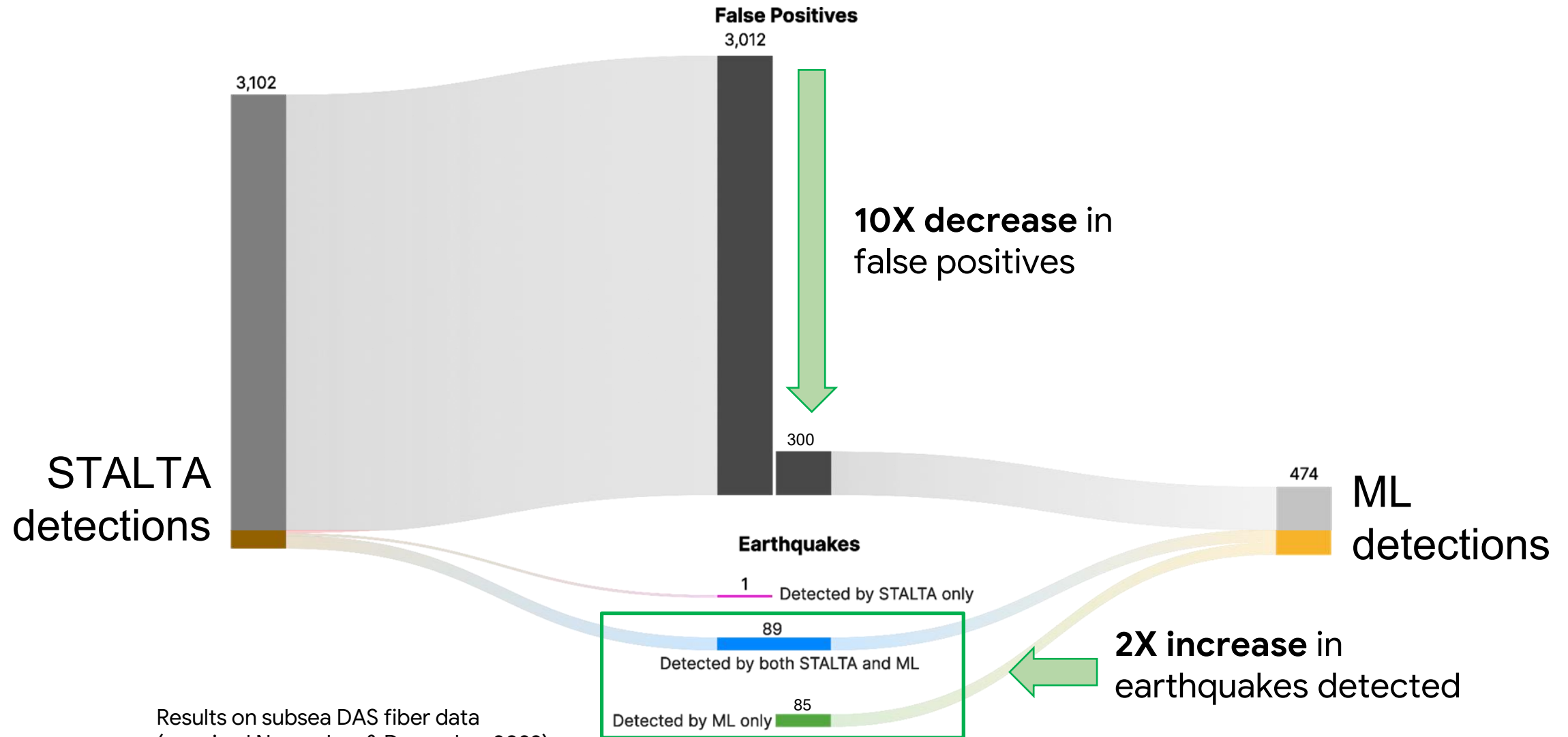


ML MODEL INFERENCE IN ACTION

- Two-minute time windows are divided into patches
- Each patch is evaluated by the ML model (earthquake, anomaly, background)
- Post-process the detection patches
- Derive detection curve from the patches

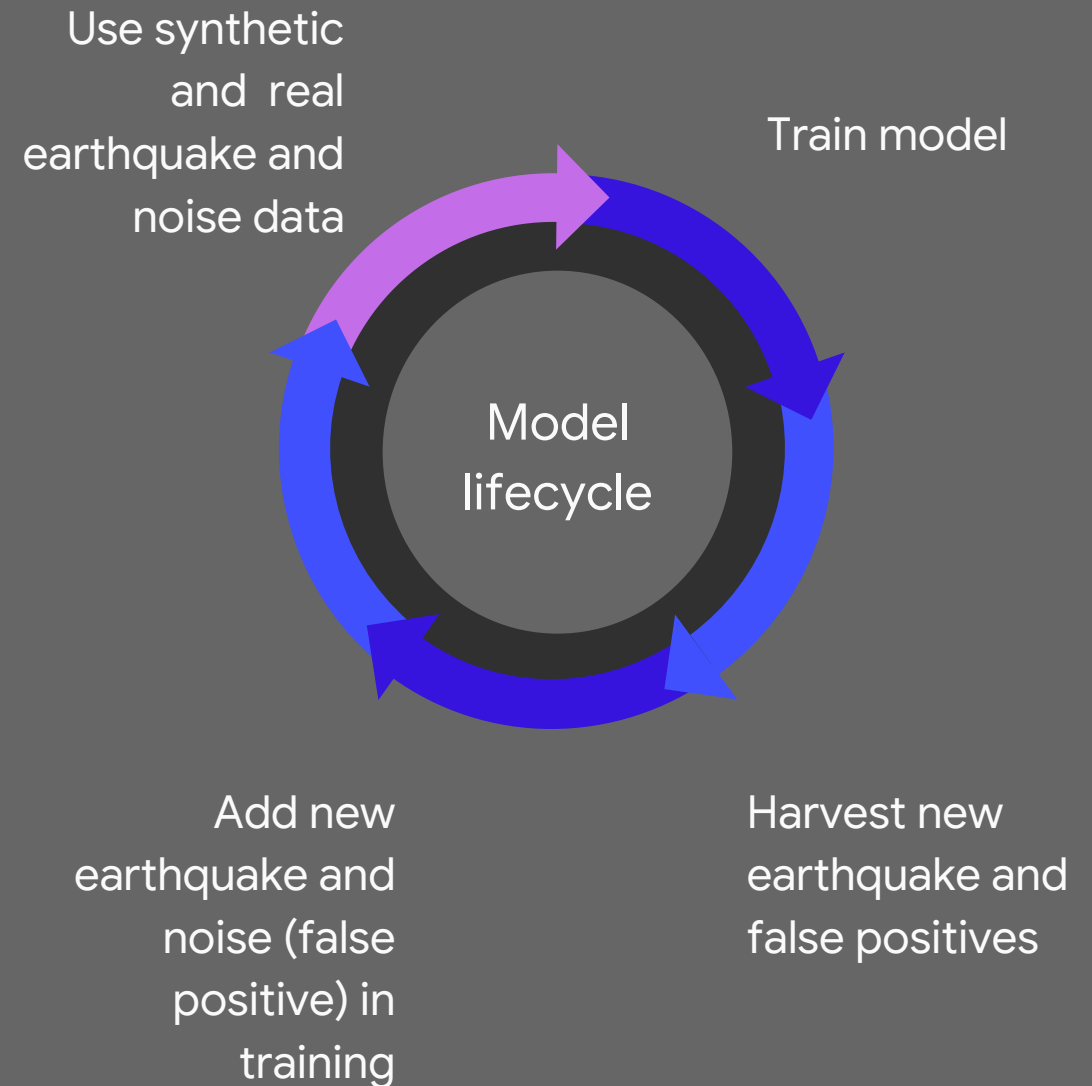


Compared to STALTA, the ML system has 10X fewer false detections while identifying 2X more earthquakes



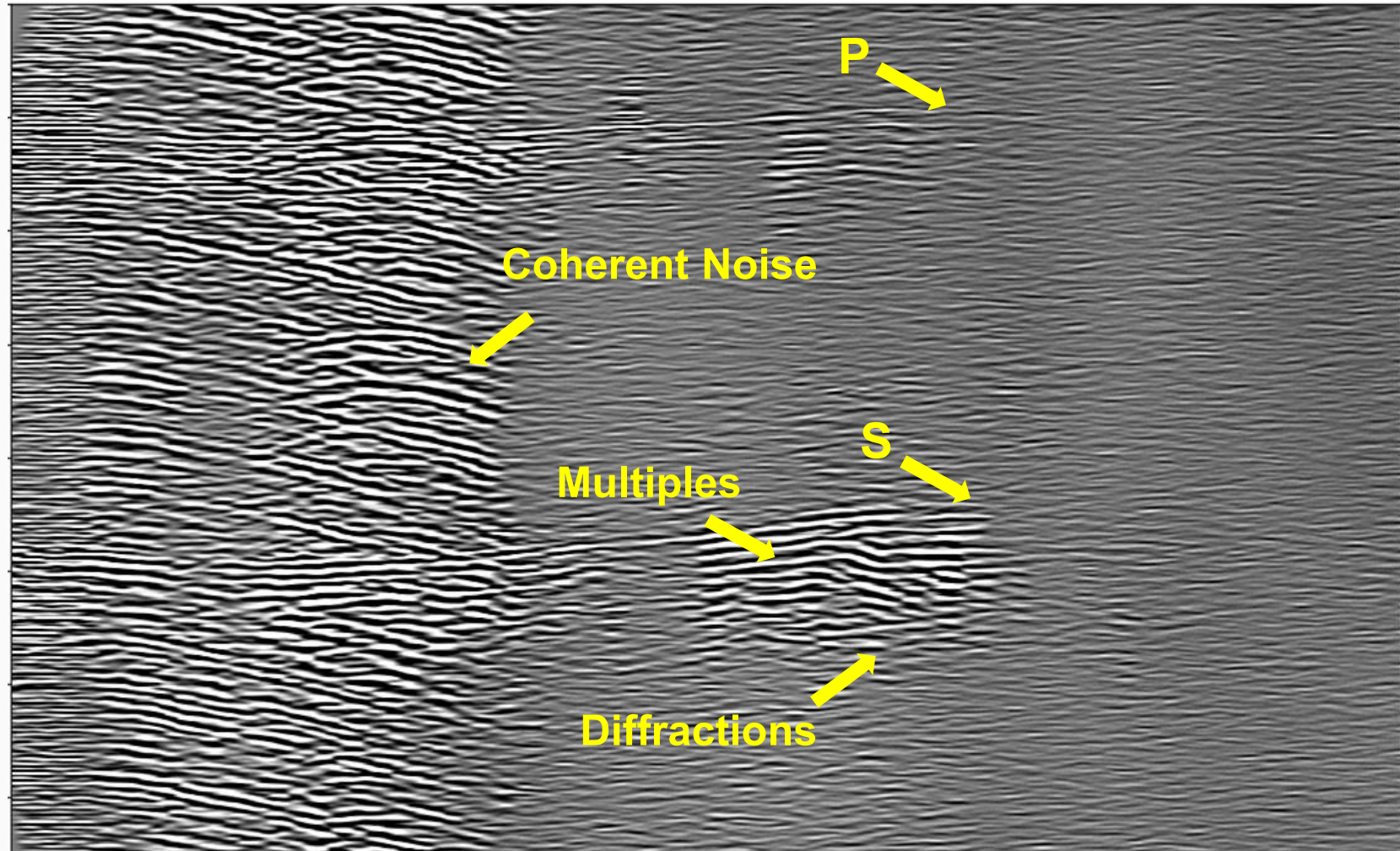
ML Flywheel: Infrastructure to wrangle terabytes of data is imperative

- Manual and ML-assisted labeling tools to facilitate initial training / active learning
- Synthetic data generation to bootstrap detector / PS picker training and augmentation
- Ability to run models on large amounts of continuous data, not just the existing event catalog (if it exists) to assess performance and harvest fainter events
- ML experiment tracking
- Hyperparameter tuning
- Parallelization and autoscaling



SYNTHETICS AS A DATA BOOTSTRAP AND MODEL PERFORMANCE BOOSTER

- Cheap to generate thousands of examples.
- Bootstrapping ML models when real events are scarce
- Augment real events to cover broader magnitude and hypocenter ranges.
- Synthetics need to look realistic. Otherwise ML models don't work.



Thank you



X, The
Moonshot
Factory