

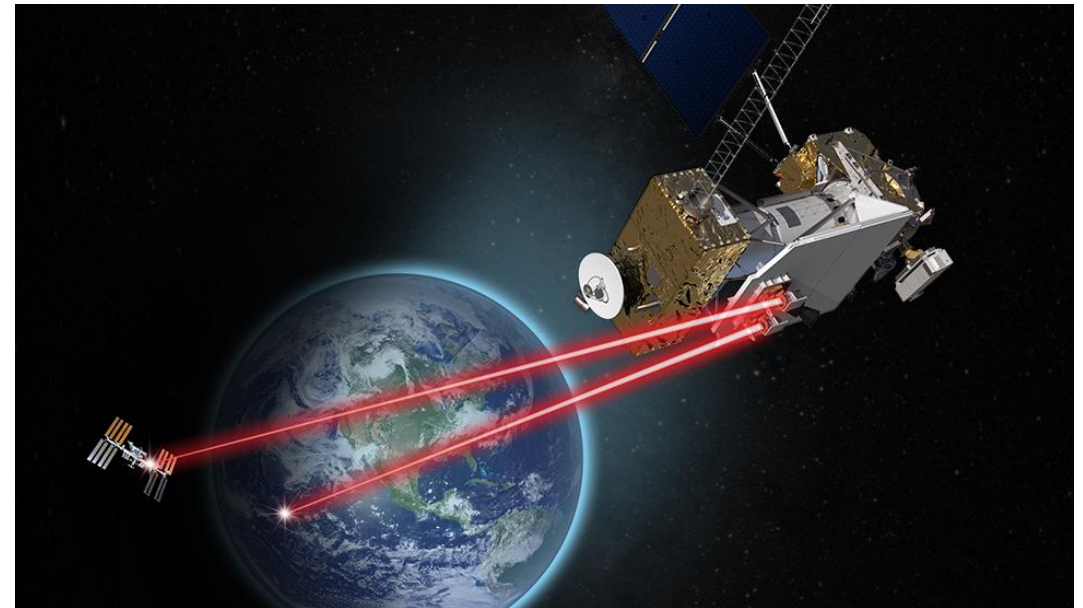
# Pointing Error And Mitigation Techniques Using Machine Learning Algorithms in Free Space Optical Communication

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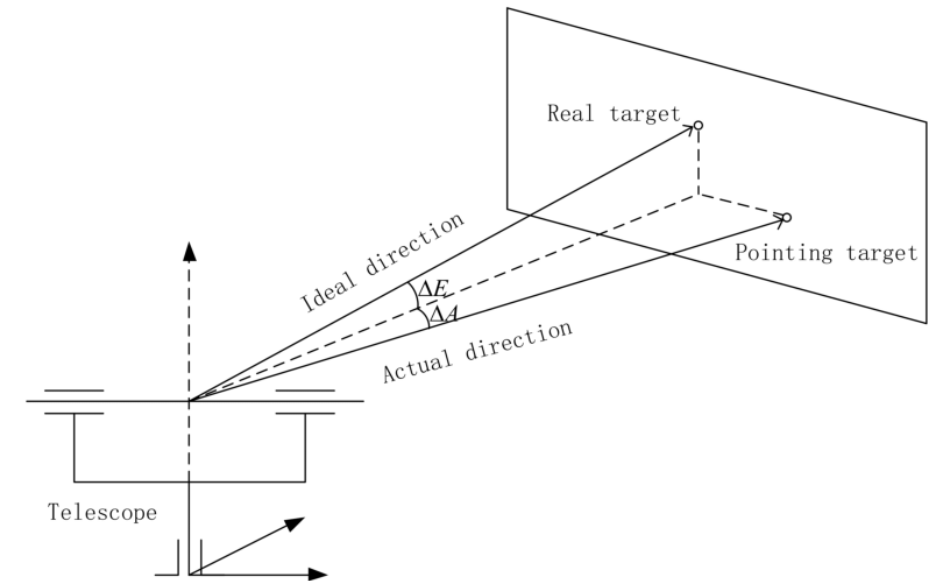
# Introduction: Free Space Optical Communication

- FSO communication uses light to transmit data between satellites or ground stations.
- **Applications** : Satellite networks (e.g., Starlink), deep-space missions (NASA), military and disaster recovery operations.
- **Advantages** : High Bandwidth, low power, quick deployment.



# Introduction: Pointing Errors

- Misalignment between the transmitted beam and the receiver aperture.
- Signal power loss, increased bit error rates, potential link failure
- Mechanical vibrations, Point-Ahead Angle (PAA) misalignment, Acquisition, Pointing, and Tracking (APT) errors, Beam divergence over long distances.



# Mitigation Techniques: Traditional Methods

- **Mechanical Systems:**
  - **Fast-Steering Mirrors (FSMs):** Small, motor-driven mirrors that adjust rapidly to correct beam direction. FSMs help compensate for small, rapid disturbances (e.g., platform jitter).
  - **Gimbals:** Larger mechanical structures that rotate entire optical components to adjust beam alignment. Useful for larger corrections but slower than FSMs.
  - **Inertial Measurement Units (IMUs):** Sensors that detect angular velocity and accelerations to stabilize and correct the beam's pointing direction.
- **Adaptive optics:** Uses deformable mirrors that change shape to counteract the effects of atmospheric turbulence.
- **Feedback Control Systems:** Systems like Proportional-Integral-Derivative (PID) controllers adjust the beam alignment by continuously monitoring errors and applying corrective signals.
- **Wavelength Division Multiplexing (WDM):** A technique that splits the optical signal into multiple wavelengths (colors of light), allowing parallel transmission of data streams.

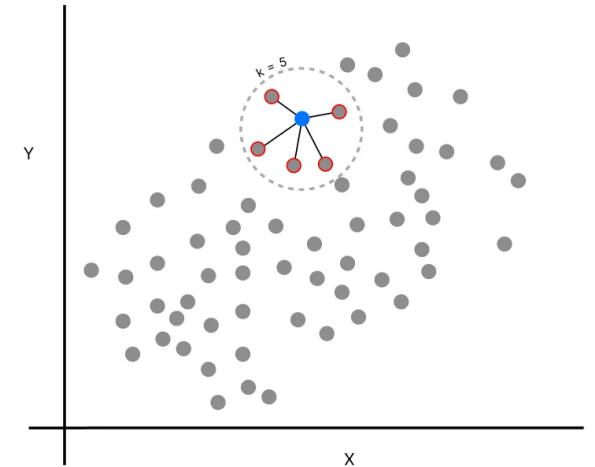
# Mitigation Techniques: Machine Learning Approach

- **Why Machine Learning models?**
  - Real time correction capabilities
  - Ability to handle non linear dynamics
  - Availability of cheap computational resources.
- **ML Models:**
  - K-Nearest neighbor(KNN)
  - Tree-based Regression Model
  - Convolution Neural Networks(Conv1D)



# K-Nearest Neighbor (KNN) Approach

- KNN is a nonparametric machine learning algorithm that predicts a value based on nearby (or "neighboring") data points.
- The model helps reduce residual nonlinear errors that traditional mitigation techniques fail to address. It achieves this by learning from historical calibration data (e.g., known star positions and pointing errors).
- **Steps:**
  - *Distance Measurement*
  - *neighbor Selection using Generalized cross Validation*
  - *Gaussian Based Weight Calculation*
  - *Final Prediction*
- **Results:** After parameter model correction, the pointing error was reduced to  $87.3 \mu\text{rad}$ . Finally, after KNN correction, the pointing error was further reduced to  $69.0 \mu\text{rad}$  for calibration stars and  $70.8 \mu\text{rad}$  for target stars.



# Tree-Based Regression Model Approach

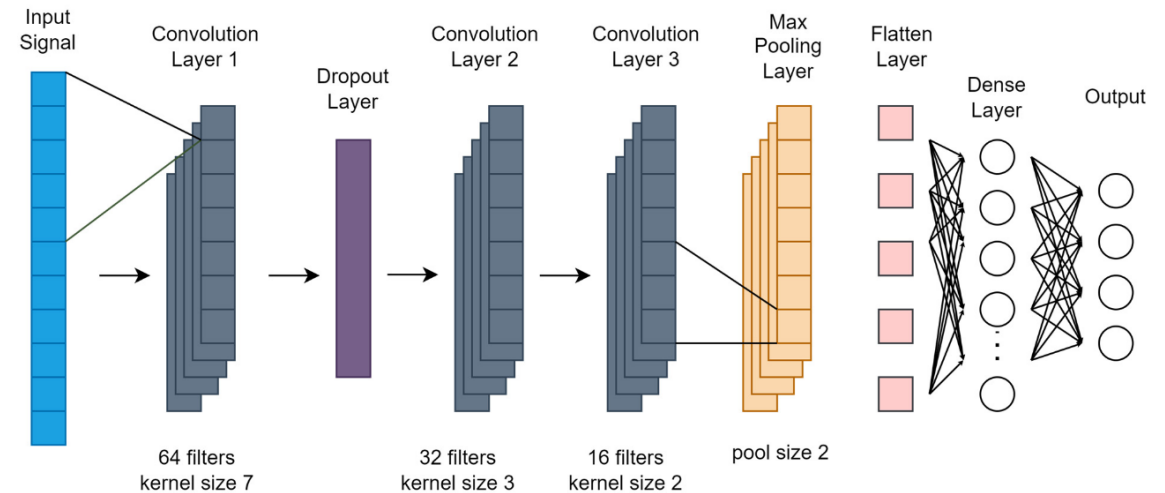
- A multi-output regression model predicts and optimizes the **gain matrix K** (multi output) to minimize pointing error.
- Models : *Decision Tree, Random Forest and Gradient Boosting Regression*
- **Steps:**
  - *Feature Extraction (System matrices, control matrix, Noise parameters)*
  - *Dataset trained by the three models*
  - *Performance matrices Evaluation(MSE, MAE, RMSE)*
  - *Gain matrix prediction*
- **Result:** Across all the performance matrices *Decision Tree* performed well compared to Random Forest and Gradient Boosting Regression.

# Convolutional Neural Network (Conv1D) Approach

- **Conv1D** is a 1-dimensional version of CNN that works well with time-series data by analyzing sequential relationships over time.
- The model's output is fed into a **closed-loop control system**, which continuously adjusts the transmitter's position to maintain beam alignment.

- **Steps:**

- *Feature Extraction*
- *Activation Function Selection (eg ReLU)*
- *Pooling for Dimensionality Reduction*
- *Regularization : dropout rate 0.5*
- *Dense Layer creation*
- *Gain Matrix prediction and closed-Loop feedback*



- **Result:** the Conv1D model achieved an **R-Squared score of 0.968**, which is higher than other tree-based regressors, indicating that Conv1D is more effective in predicting the gain matrix



# Conclusion

- Each model has its unique advantages depending on the complexity of the system and the nature of the data.
- **K-Nearest neighbor (KNN):** This algorithm effectively corrects nonlinear pointing errors and improves pointing accuracy.
- **Tree-based regression models:** The models were tested on open-loop and closed-loop systems, and the results showed that the tree-based regression model effectively mitigates pointing errors in FSO systems.
- **Convolutional Neural Network:** By continuously feeding real-time pointing error measurements into the model, it dynamically adjusts the transmitter's position to maintain beam alignment with the receiver.

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# Thank You

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