

# **Autonomous system numbers based internet censorship detection**

**Using machine learning to identify network disruptions**

# Project Overview

## Goal

- Detect internet potential censorship events using Autonomous System Number (ASN) data
- Apply both supervised and unsupervised machine learning approaches

## Data Sources

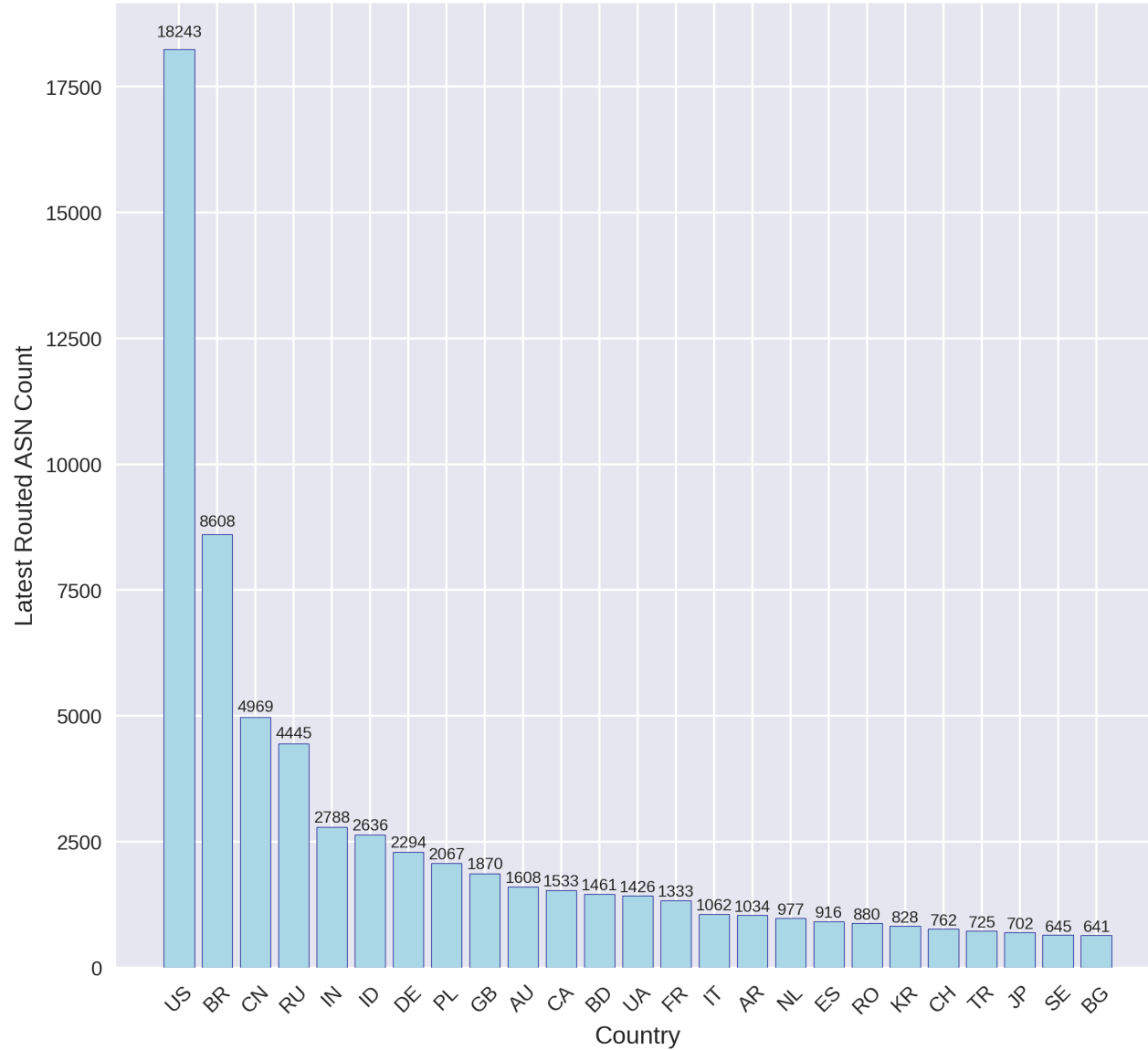
- RIPE API <https://www.ripe.net/>
  - RIPE NCC is the regional Internet registry for Europe, the Middle East, and parts of Central Asia



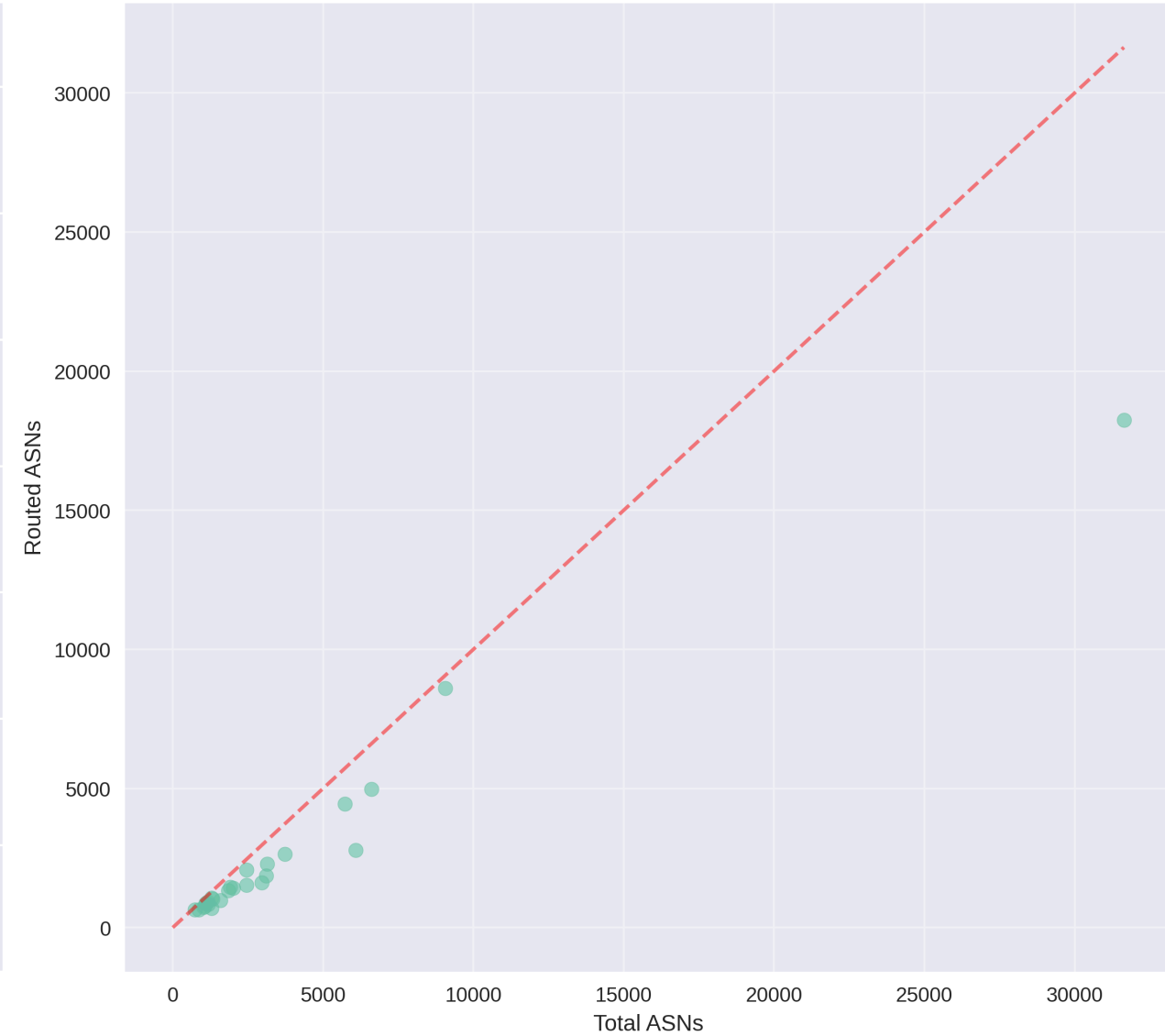
- An autonomous system (AS) is a very large network or group of networks with a single routing policy.
- In a simplified case it is a network of a single internet provider or company serving traffic (e.g. AWS and Cloudflare each has several ASs at least).
- AS can be routed (serve traffic), or only registered (not in work)
- AS can be connected to other inland ASs or to foreign ASs

## Latest ASN Counts - Snapshot View

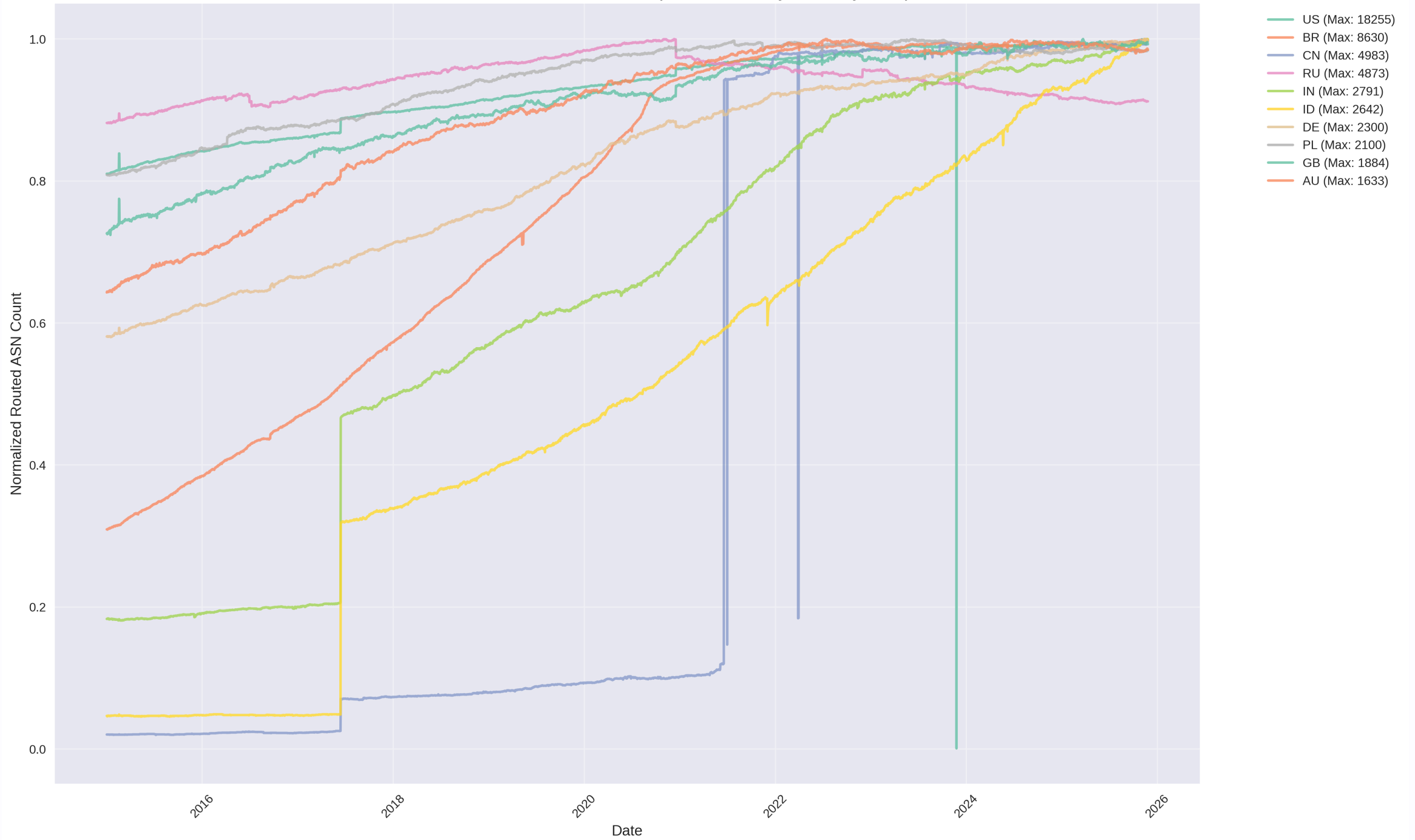
Top 25 Countries by Latest Routed ASN Count



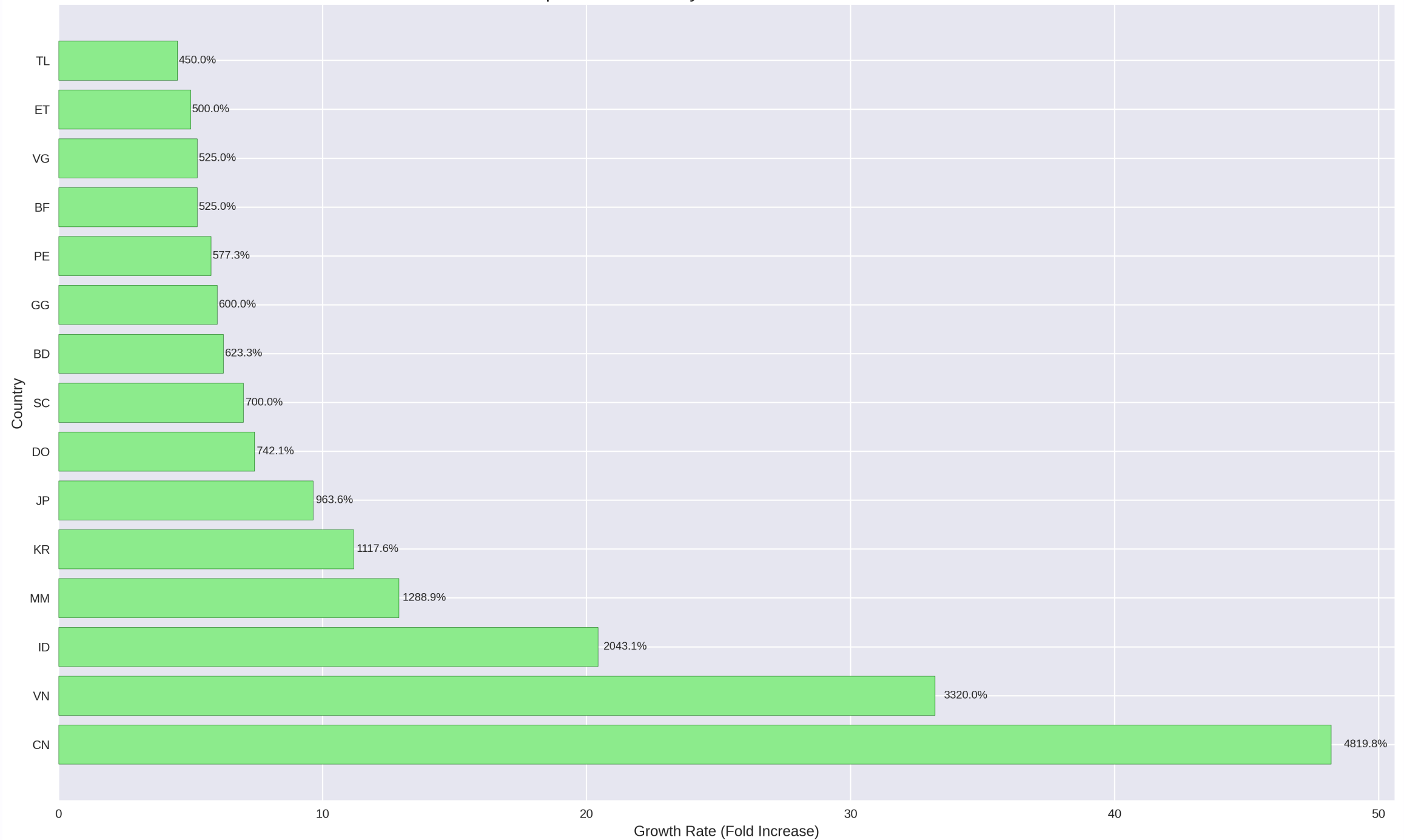
Routed vs Total ASNs (Latest Values)



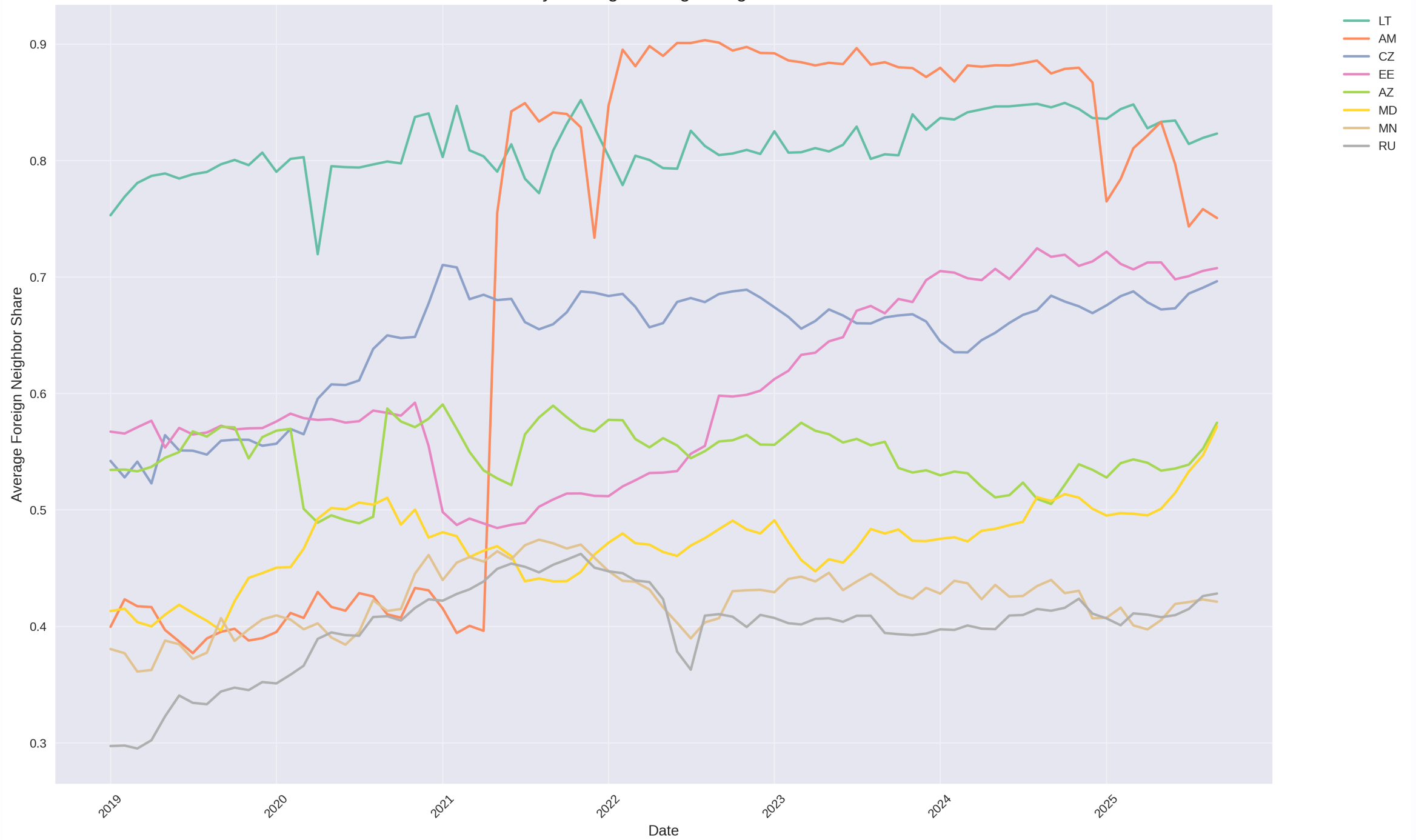
Time Series of Routed ASN Counts (Normalized by Country Max)



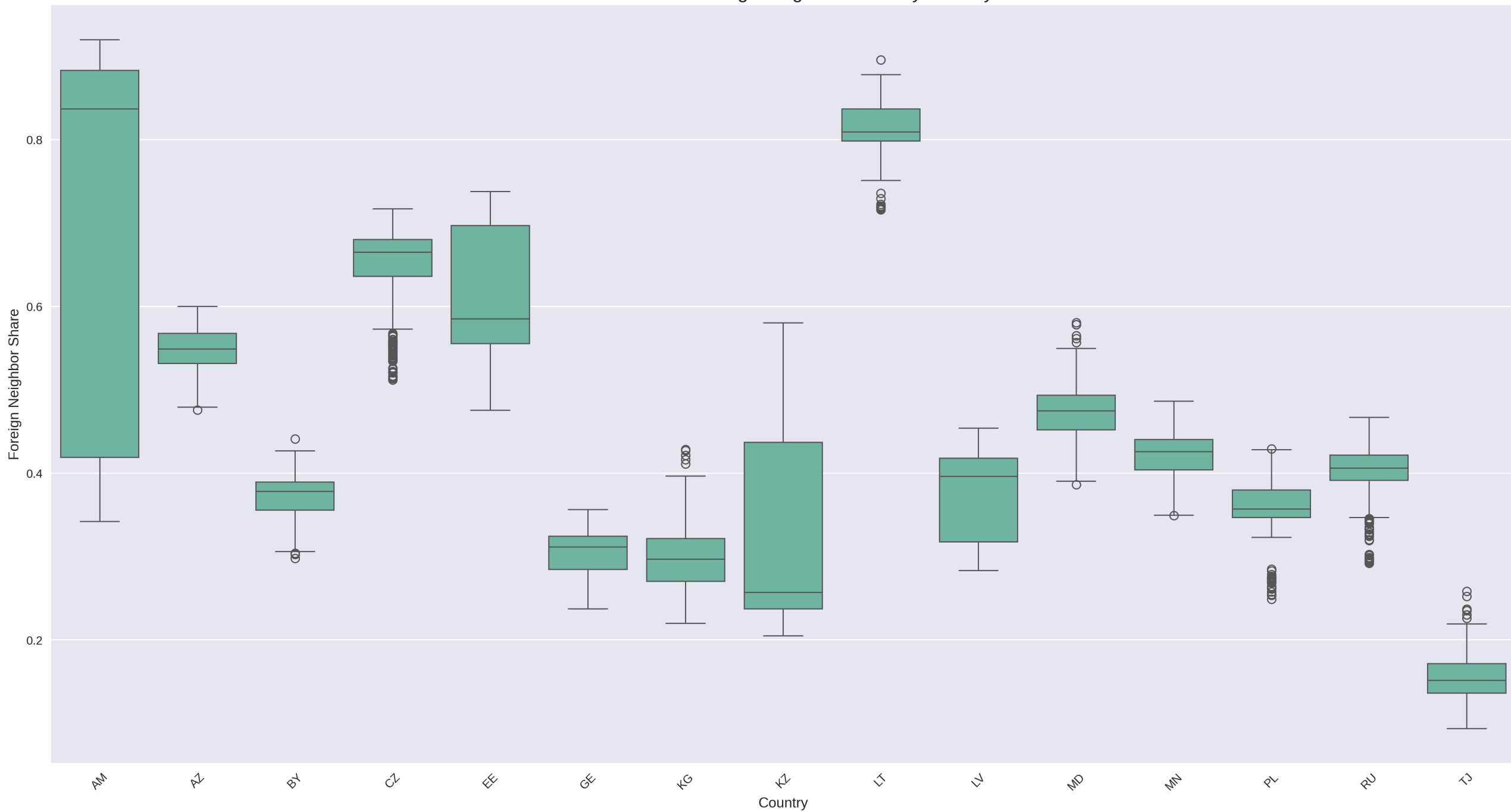
Top 15 Countries by ASN Routed Growth Rate



Monthly Average Foreign Neighbor Share



Distribution of Foreign Neighbor Share by Country





# Machine Learning

## Feature Engineering

- Created `asn_censorship_indicator` based on significant ASN count drops
- Calculated connectivity metrics including foreign neighbor share

## Models

- Logistic Regression
- Random Forest Classifier
- Time Series Autoencoder for anomaly detection

Layer (type)	Output Shape	Param #
<code>lstm (LSTM)</code>	<code>(None, 30, 64)</code>	17,152
<code>lstm_1 (LSTM)</code>	<code>(None, 32)</code>	12,416
<code>repeat_vector (RepeatVector)</code>	<code>(None, 30, 32)</code>	0
<code>lstm_2 (LSTM)</code>	<code>(None, 30, 32)</code>	8,320
<code>lstm_3 (LSTM)</code>	<code>(None, 30, 64)</code>	24,832
<code>time_distributed (TimeDistributed)</code>	<code>(None, 30, 2)</code>	130

# Key Findings

## Potential censorship events detected

- **Total Events:** 1,367 potential censorship events
- **Affected Countries:** 75 countries
- **Top Country:** Dominica (DM) with 289 events

## Top Countries

1. Dominica (DM) - 289 events
2. Syria (SY) - Significant events
3. Saint Kitts and Nevis (KN) - Significant events
4. Suriname (SR) - Significant events
5. Kiribati (KI) - Significant events

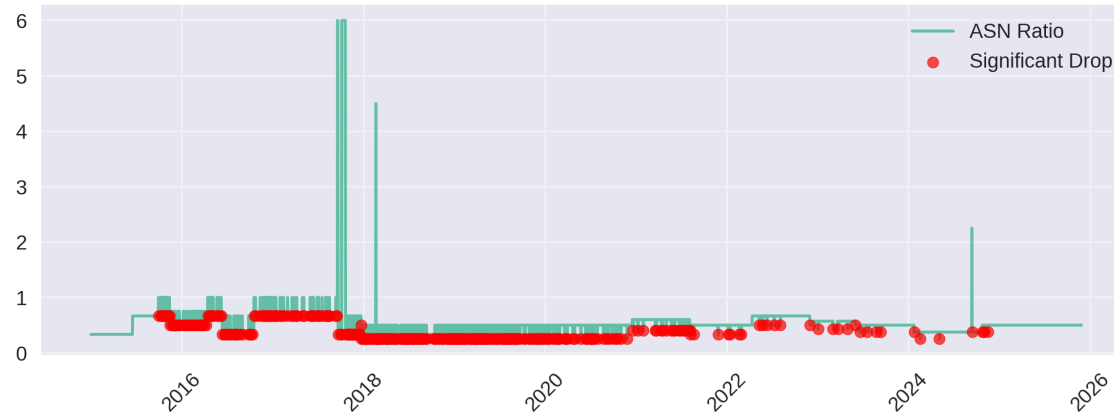
# Supervised vs Unsupervised Comparison

## Method Comparison

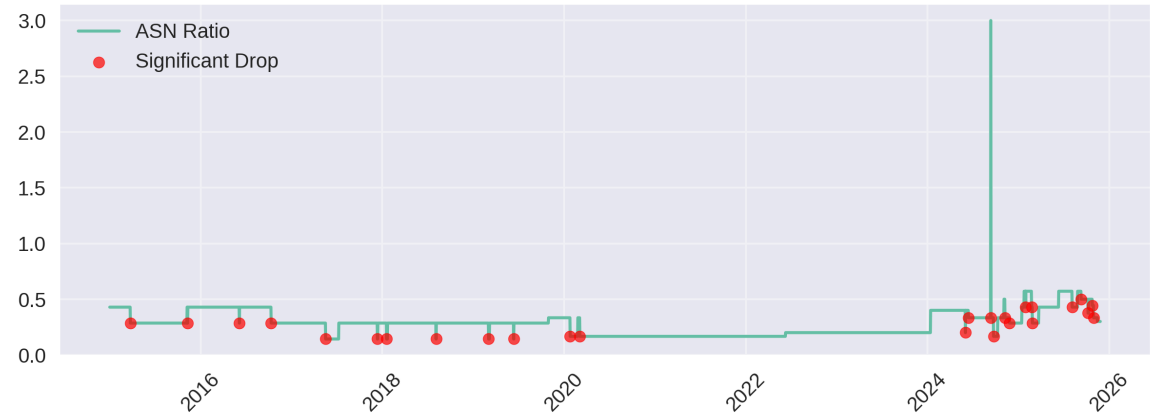
- *Supervised*: 1,367 events detected using labeled patterns
- *Unsupervised*: 87 anomalies detected using statistical methods
- *Overlap*: 63 events detected by both methods (4.61% overlap)

# ASN Ratio Changes - Identifying Potential Censorship Events

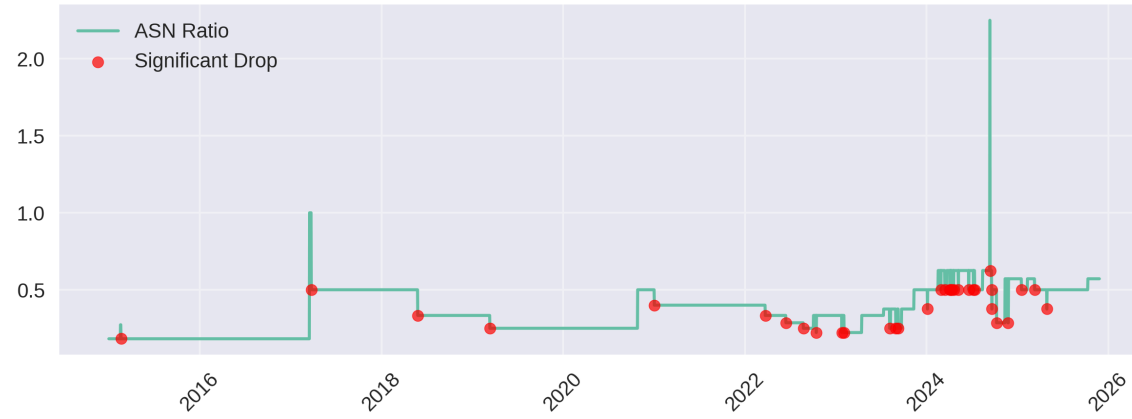
## ASN Ratio Trend - DM



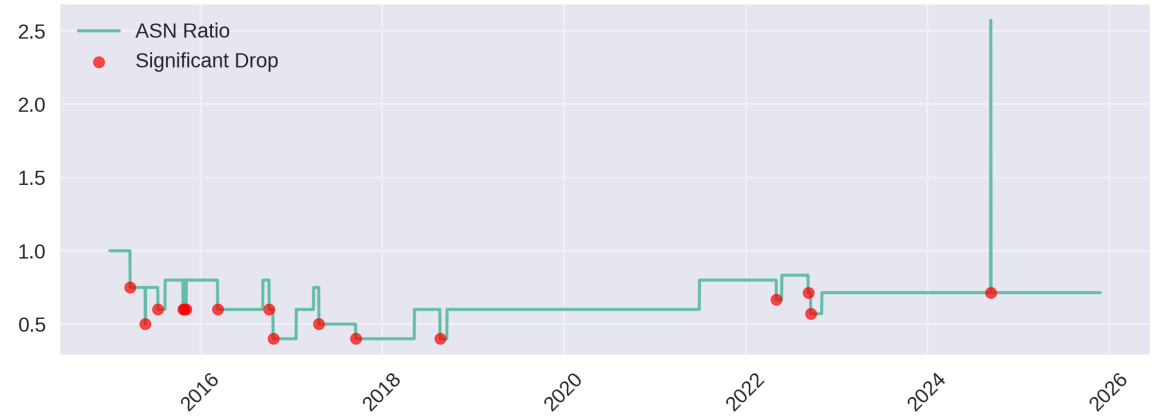
## ASN Ratio Trend - SY



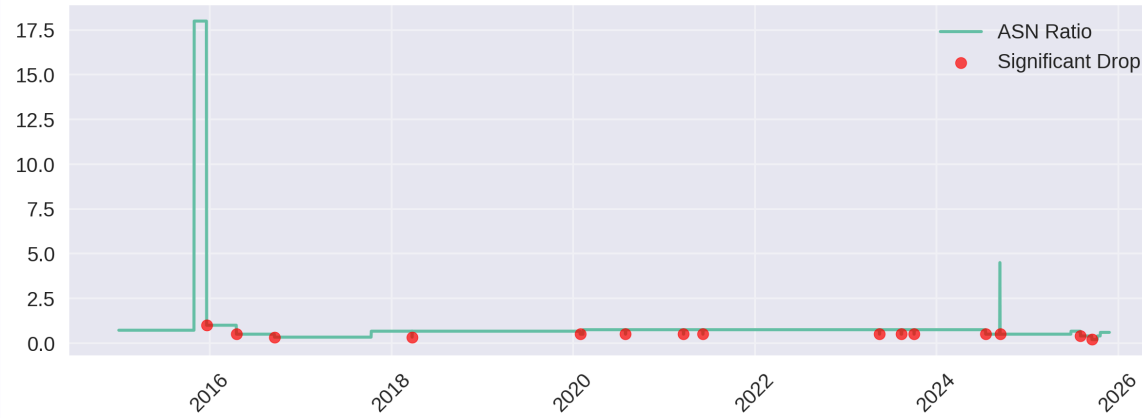
## ASN Ratio Trend - KN



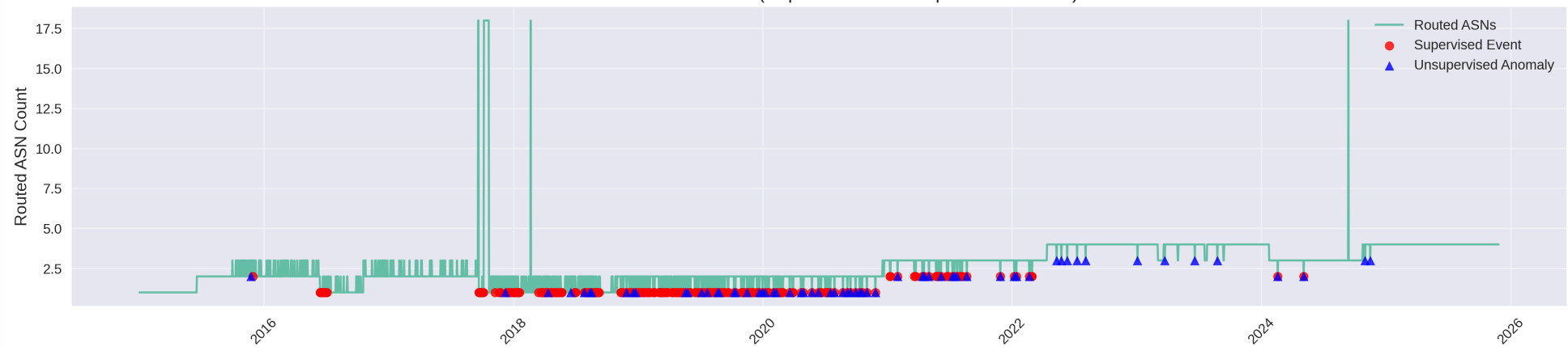
## ASN Ratio Trend - SR



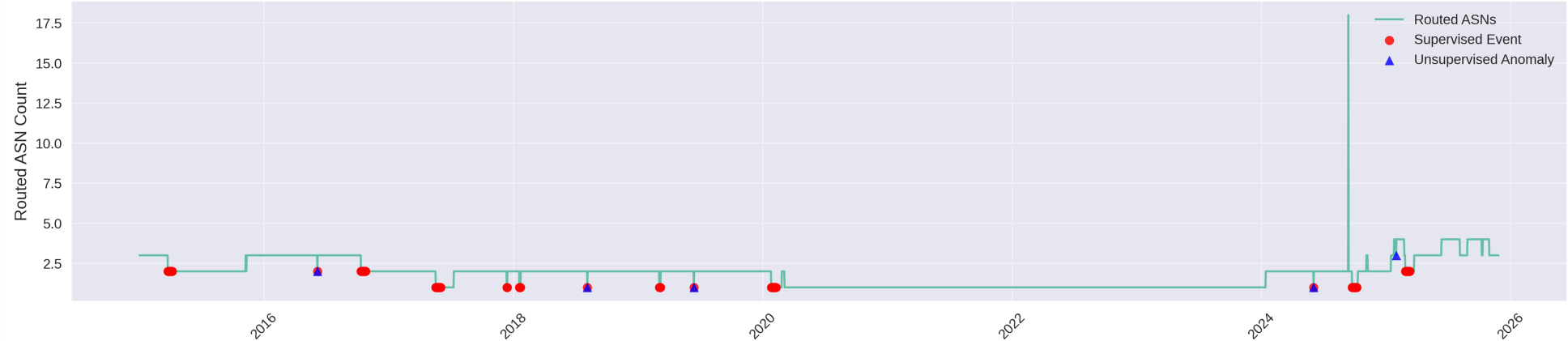
## ASN Ratio Trend - KI



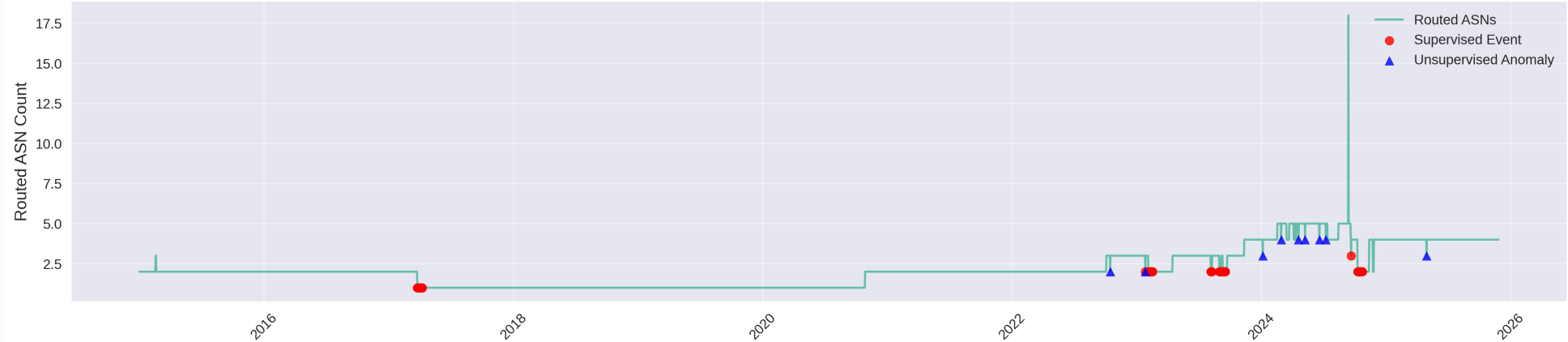
Comparison of Supervised and Unsupervised Event Detection  
ASN Time Series - DM (Supervised vs Unsupervised Events)



ASN Time Series - SY (Supervised vs Unsupervised Events)



ASN Time Series - KN (Supervised vs Unsupervised Events)



# Results Summary

## Key Insights

- Different detection methods identify complementary event patterns
- ASN data provides valuable insights into network disruptions
- Many significant drop in ASN are small numbers effect

# Conclusion

## Future Work

- Handle small countries
- Cross-reference with known censorship events
- Refine unsupervised anomaly detection algorithms
- Apply to real-time monitoring systems