

RISK AND RESILIENCE BOOTCAMP





WORKFORCE
DEVELOPMENT

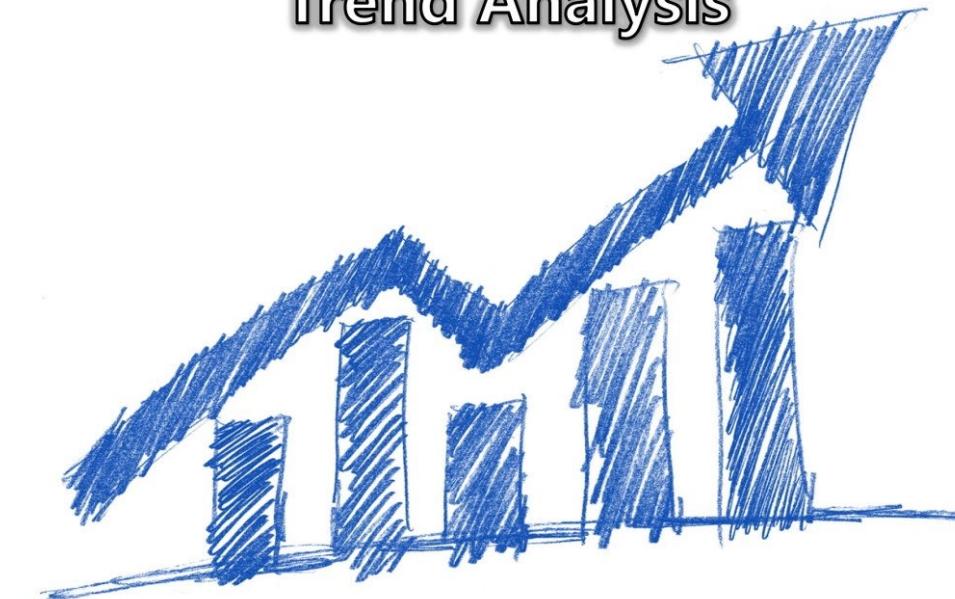


RISK TREND AND AGGREGATES

This module is an introduction to some analytics used in risk monitoring

- Trend Analysis
- Aggregation Analysis

Trend Analysis



iEduNote.com

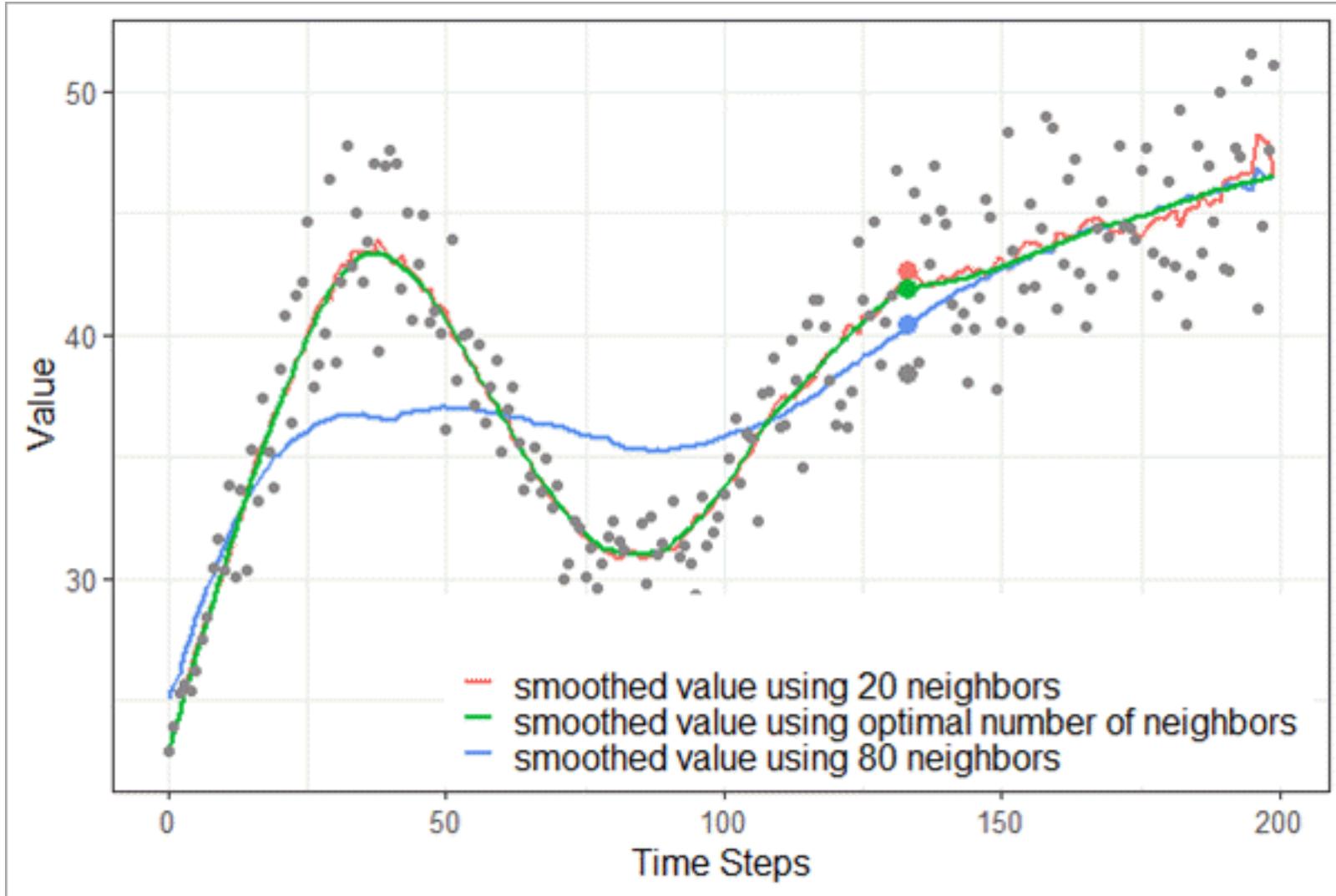
TIME SERIES/AGGREGATE ANALYSIS

- Monitoring longitudinal behavior, not just point-in-time metrics
 - Time-series approaches detect patterns, anomalies, and accumulative risks that indicate systemic vulnerabilities
 - Why time-series analysis matters
 - Identifies slow-moving, chronic failures (e.g., growing patch backlog).
 - Reveals seasonality or cyclical patterns (e.g., transaction spikes causing stress).
 - Highlights drift in performance or control effectiveness.
 - Provides early warnings when multiple KRIs deteriorate simultaneously.
 - Mature organizations aggregate metrics across:
 - Business units (e.g., rising access exceptions across departments)
 - Systems (e.g., API latency trends across microservices)
 - Geographies (e.g., region-wide authentication failures)
 - Control families (e.g., decline in detective control success rates)

ADVANCED ANALYTICAL TECHNIQUES

- Moving averages /exponential smoothing
 - Purpose
 - Smooth noisy data to expose underlying trends and direction.
 - Example in risk monitoring
 - KRI: Failed login attempts per hour (indicator of brute-force attack or credential abuse)
 - Raw failed login counts are highly volatile
 - Using a 7-day moving average, a rising trend becomes visible even if daily values fluctuate
 - Exponential smoothing assigns higher weight to recent values, making emerging attacks easier to spot
 - Outcome
 - The SOC detects an upward directional trend in authentication anomalies
 - This triggers proactive controls like throttling, geo-blocking, or requiring MFA challenges

SMOOTHING

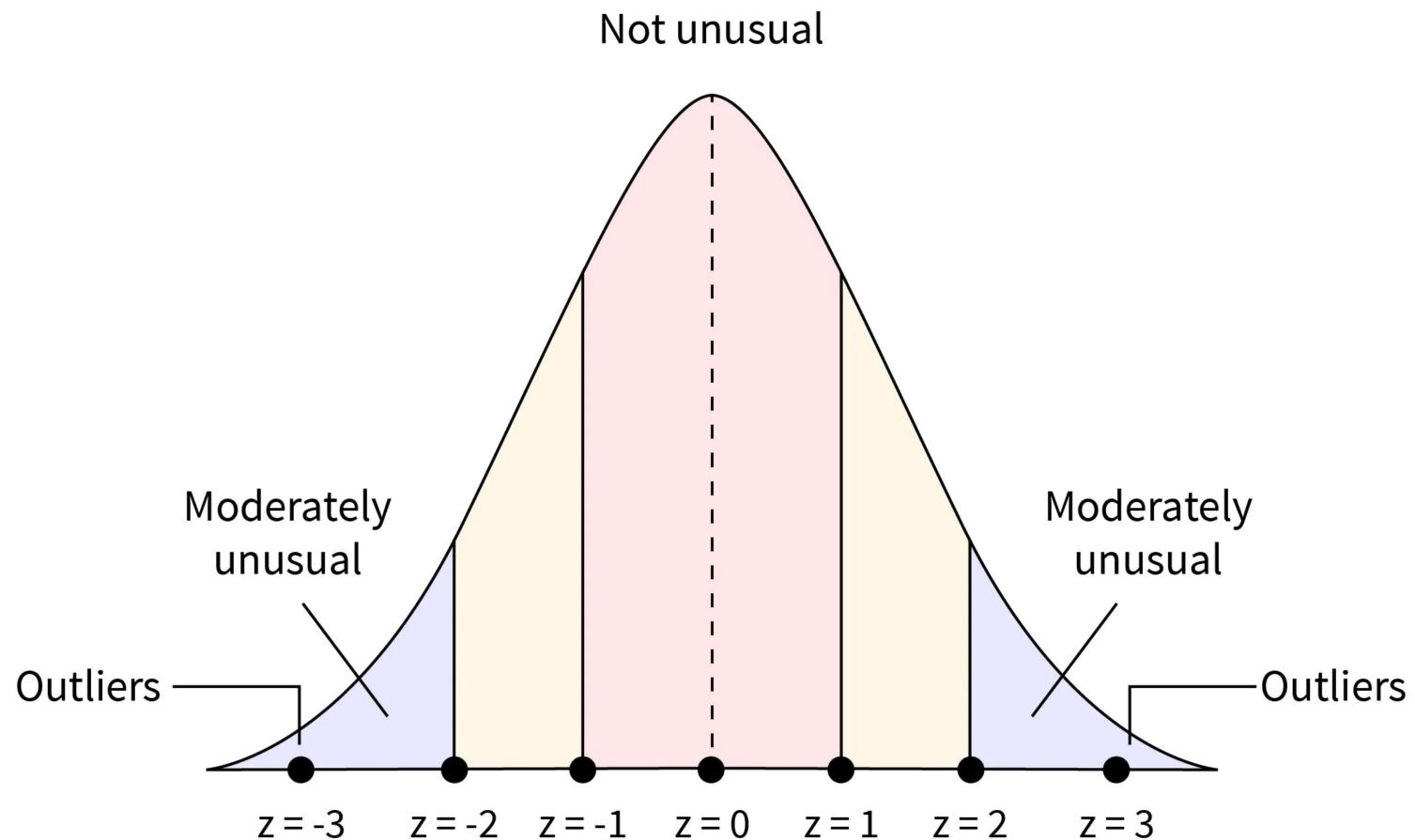


ADVANCED ANALYTICAL TECHNIQUES

- Z-Score Normalization / outlier detection
 - Purpose
 - Identify values that deviate significantly from normal behavior
 - Example in risk monitoring
 - KRI: Outbound data transfer volume from a sensitive database
 - The system computes a Z-score for each hour's outbound traffic
 - Z-score > 3 (three standard deviations above the mean) triggers a security alert for potential data exfiltration
 - Even if the transfer size is below an absolute threshold, statistical deviation flags the anomaly
 - Outcome
 - An insider threat or compromised account can be detected before significant data is lost, based solely on abnormal behavior relative to historical patterns

Z-SCORE

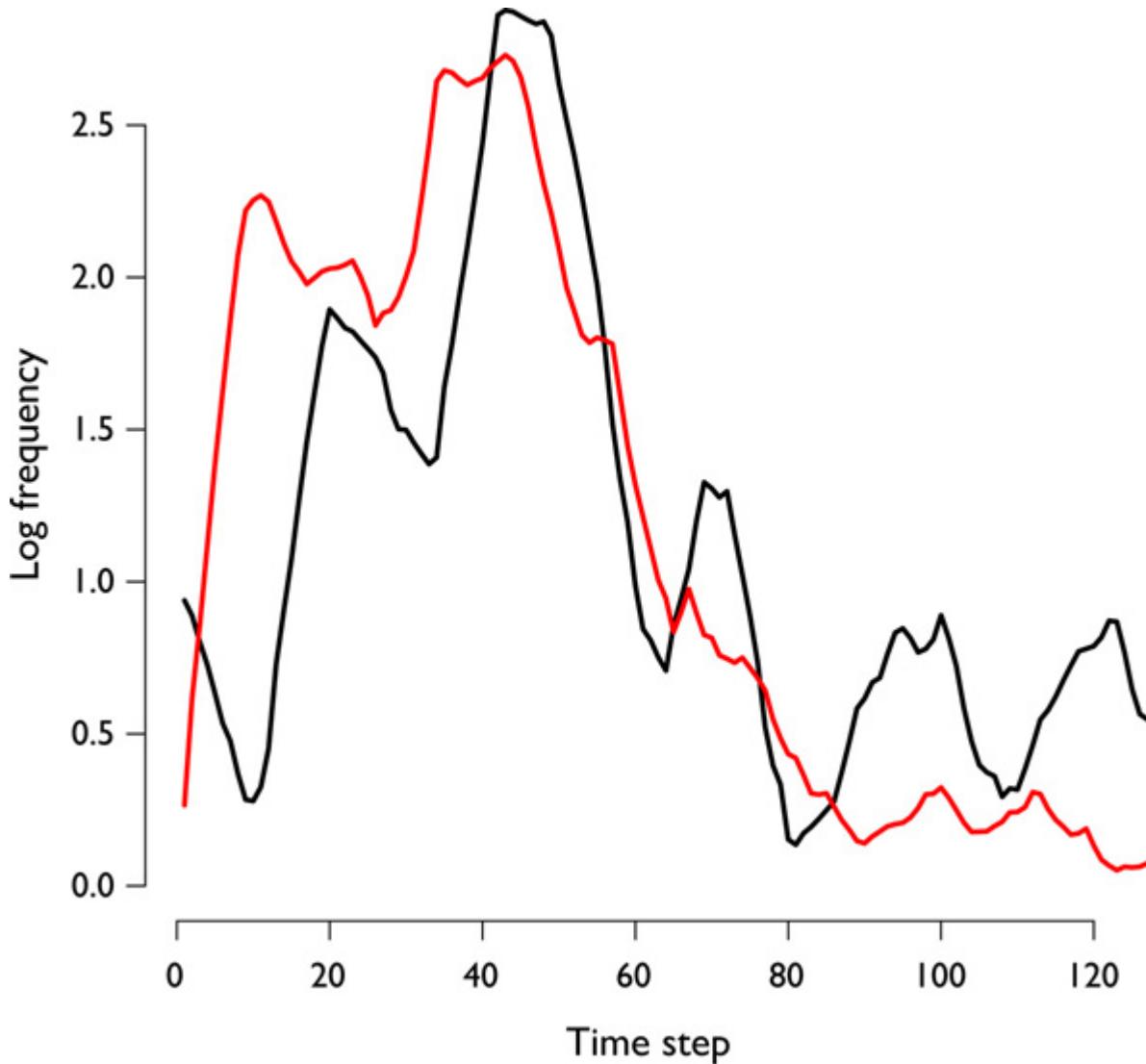
Detecting Outliers with z-Scores



ADVANCED ANALYTICAL TECHNIQUES

- Lag analysis
 - Purpose
 - Measure whether changes in one metric precede changes in another metric
 - Example in risk monitoring
 - KPI: Patch completion rate vs KRI: Count of critical vulnerabilities
 - Lag-7 analysis reveals
 - A decrease in patch completion today correlates with
 - An increase in critical vulnerabilities seven days later
 - Outcome
 - Risk teams can demonstrate a causal or predictive relationship between a KIR and heightened vulnerability exposure which strengthens the case for resource allocation or process redesign

LAG ANALYSIS

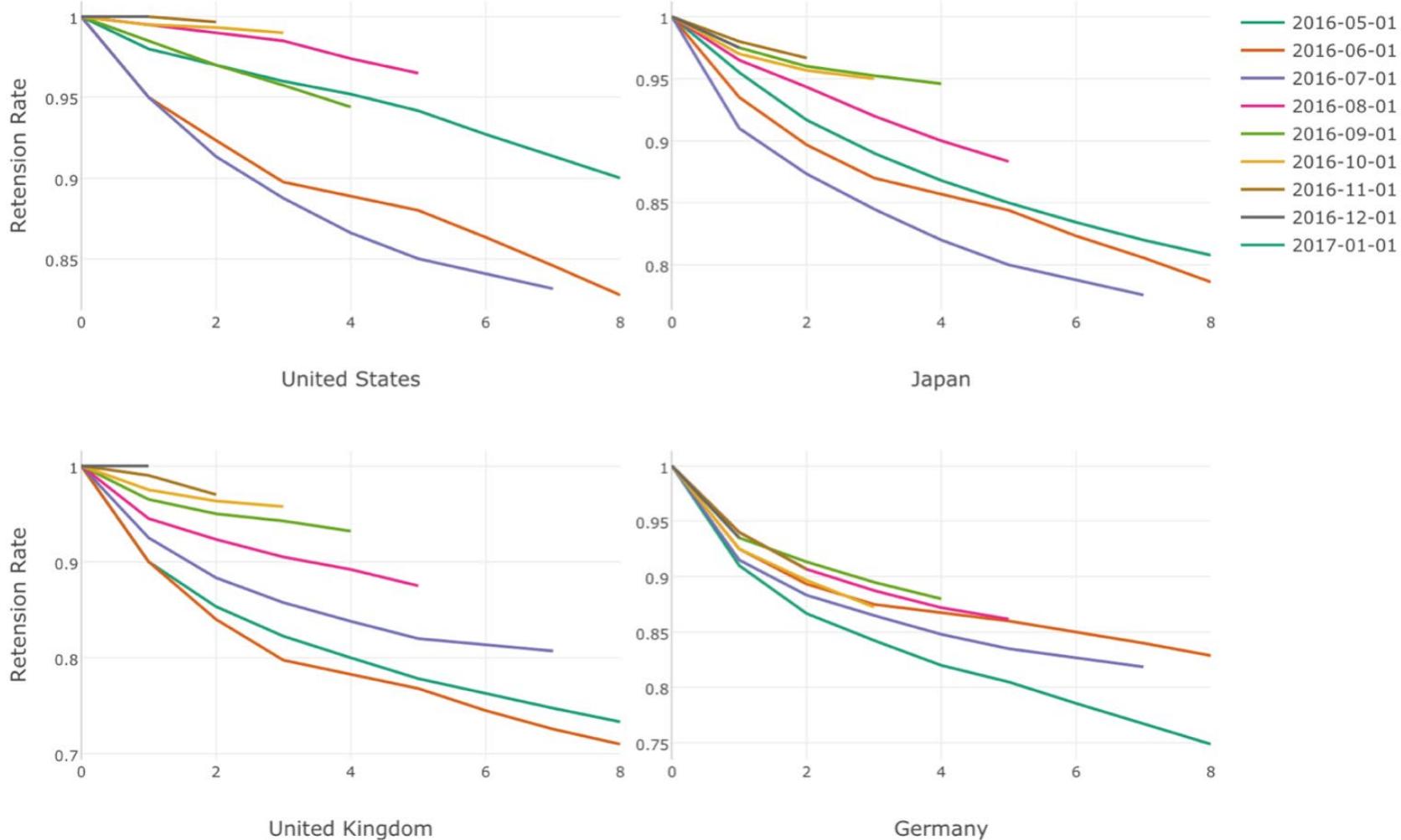


ADVANCED ANALYTICAL TECHNIQUES

- Cohort trend analysis
 - Purpose
 - Compare similar groups (cohorts) to find patterns or outliers within a category
 - Example in risk monitoring, cohorts are created by
 - Application type (web apps vs. batch apps)
 - Technology stack (Java vs. Node vs. Python)
 - Environment (production vs. staging)
 - Ownership (Team A vs. Team B)
 - A cohort analysis reveals
 - Systems owned by Team B show a 3× higher growth rate of high-severity vulnerabilities compared to similar systems owned by other teams.
 - Outcome
 - This reveals a team-specific process breakdown rather than a global security weakness suggests targeted training, staffing, or process intervention can be deployed

COHORT ANALYSIS

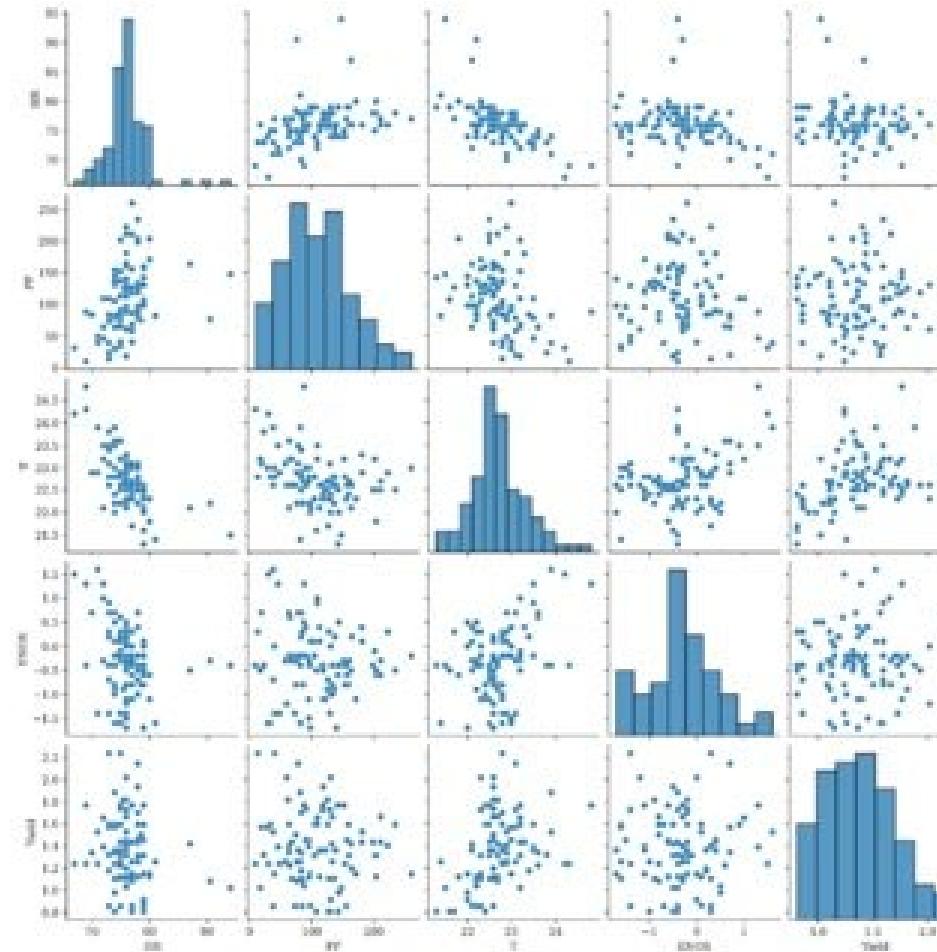
Customer Retention Rates by Month and Country



ADVANCED ANALYTICAL TECHNIQUES

- Cross-correlation matrices
 - Purpose
 - Identify how metrics influence each other across time.
 - Metrics include
 - KPI: Mean Time to Recover (MTTR) for incidents
 - KRI: Incident frequency
 - KPI: Change success rate
 - KRI: Number of repeated near misses
 - Cross-correlation shows
 - When change success rate decreases
 - Incident frequency increases with a correlation coefficient of 0.82
 - Additionally, higher MTTR correlates with a rise in repeated near misses, suggesting control inefficiency
 - Outcome
 - The correlation matrix grounds the argument that unstable change processes are the root cause of operational risk increases which should trigger major change-control reforms

CROSS-CORRELATION MATRICES



ADVANCED ANALYTICAL TECHNIQUES

- Composite KRI Index
 - Purpose
 - Combine disparate KRIs into a single normalized risk score that can be tracked over time
 - Combine KRIs such as
 - Number of critical vulnerabilities (weighted 30%)
 - Privileged access exceptions (20%)
 - Failed login anomalies (20%)
 - Backup integrity failures (15%)
 - End-of-life infrastructure ratio (15%)
 - Outcome
 - The correlation matrix grounds the argument that unstable change processes are the root cause of operational risk increases which should trigger major change-control reforms

ADVANCED ANALYTICAL TECHNIQUES

- Composite KRI Index
 - Using normalized values and weights produces a Composite KRI Index (0–100).
 - Example trend
 - Q1: 42
 - Q2: 49
 - Q3: 63
 - Q4: 71
 - Even without a major incident, the index shows a clear erosion of risk posture, prompting leadership to invest in remediation before conditions worsen
 - Composite indices provide:
 - Board-friendly reporting
 - Cohesive trend visibility
 - A single metric that captures complex, multidimensional risk exposure
 - This enables strategic prioritization and budget planning

Q&A AND OPEN DISCUSSION

