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DESIGNING GEOPOLITICAL RESILIENCE INTO ORACLE GTM AND OTM THROUGH AI-ENABLED CUSTOMIZATION

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ABSTRACT

Global supply chains face unprecedented disruption from geopolitical tensions, trade policy volatility, regional conflicts, and shifting international alliances that challenge traditional transportation and logistics planning assumptions. Oracle Global Trade Management and Oracle Transportation Management systems serve as critical infrastructure for managing complex international supply chains, yet their standard configurations often lack the adaptive capabilities required to navigate rapidly evolving geopolitical landscapes. This paper presents a comprehensive framework for embedding geopolitical resilience into Oracle GTM and OTM platforms through artificial intelligence-enabled customization strategies. We examine how machine learning algorithms can analyze geopolitical risk indicators, predict disruption scenarios, and automatically trigger supply chain reconfigurations that maintain operational continuity despite political uncertainties. The proposed architecture integrates real-time geopolitical intelligence feeds with Oracle's logistics optimization engines, enabling dynamic route selection, supplier diversification recommendations, and compliance monitoring adapted to changing regulatory environments. AI-driven customizations include predictive sanctions screening, automated tariff impact analysis, intelligent routing around conflict zones, and machine learning models forecasting trade policy changes based on political developments. A prototype implementation demonstrates how natural language processing extracts actionable intelligence from news sources, government announcements, and policy documents to inform logistics decisions. Case studies illustrate the system's effectiveness during recent geopolitical events including trade wars, regional conflicts, and pandemic-related border closures. Results show that AI-enhanced Oracle systems reduced supply chain disruption duration by 43%, decreased emergency freight costs by 31%, and improved regulatory compliance rates to 99.2% despite volatile policy environments. This research provides practical guidance for supply chain organizations seeking to future-proof their Oracle logistics platforms against geopolitical uncertainty through intelligent system customization.

KEYWORDS: Geopolitical Risk, Supply Chain Resilience, Oracle Gtm, Oracle Otm, Artificial Intelligence, Trade Compliance, Logistics Optimization, Supply Chain Disruption

INTRODUCTION

The global business environment has entered an era of profound geopolitical uncertainty characterized by escalating trade tensions between major economies, regional conflicts disrupting established logistics corridors, rapidly evolving sanctions regimes, and fundamental questioning of the post-Cold War international order that enabled decades of supply chain globalization. Supply chain managers who once could rely on relatively stable geopolitical conditions now confront scenarios where established trade routes become impassable overnight, trusted suppliers fall under sanctions without warning, and tariff structures change faster than systems can be updated (Christopher and Holweg, 2017).

Oracle Global Trade Management and Oracle Transportation Management represent enterprise-grade platforms managing billions of dollars in international trade flows for the world's largest corporations. GTM handles trade compliance including classification, documentation, restricted party screening, and regulatory reporting across multiple jurisdictions. OTM optimizes transportation planning, execution, and settlement across complex multi-modal networks spanning global operations. Together, these systems form the technological backbone enabling sophisticated international supply chains (Sodhi and Tang, 2021).

However, the standard configurations of these powerful platforms were designed for operational efficiency in relatively stable environments rather than resilience amid geopolitical volatility. Traditional GTM implementations rely on periodic updates of trade regulations and restricted party lists that may lag dangerous developments by days or weeks. Standard OTM routing optimization focuses on cost and transit time without adequately incorporating geopolitical risk factors that could render optimized routes suddenly unviable. The rule-based nature of conventional configurations struggles to adapt quickly to unprecedented scenarios falling outside predefined parameters (Ivanov and Dolgui, 2020).



Artificial intelligence and machine learning offer transformative potential for enhancing Oracle logistics platforms with geopolitical awareness and adaptive capabilities. Machine learning models can analyze vast quantities of geopolitical intelligence from diverse sources, identifying patterns humans might miss and forecasting disruptions before they materialize. Natural language processing enables automated extraction of actionable logistics intelligence from news articles, government announcements, diplomatic communications, and policy documents published in multiple languages. Predictive analytics can estimate the probability and potential impact of various geopolitical scenarios, enabling proactive contingency planning (Baryannis et al., 2019).

Deep learning techniques applied to historical disruption data can identify leading indicators that specific types of geopolitical events will affect particular trade lanes or commodities. Reinforcement learning algorithms can optimize supply chain network design balancing efficiency against resilience, learning from simulated disruption scenarios which configurations maintain acceptable performance under various stress conditions. These AI capabilities can be integrated with Oracle platforms through strategic customization that extends native functionality while preserving upgrade paths and supportability (Choi et al., 2020).

This paper presents a comprehensive framework for designing geopolitical resilience into Oracle GTM and OTM through intelligent AI-enabled customizations. We examine the architecture required to ingest and analyze geopolitical intelligence, the algorithms most effective for different logistics decision contexts, and the integration patterns that extend Oracle platforms while maintaining system integrity. Case studies demonstrate practical applications and quantified benefits from implementations addressing real-world geopolitical challenges.

GEOPOLITICAL RISK LANDSCAPE FOR SUPPLY CHAINS

Understanding the diverse geopolitical threats facing international supply chains provides essential context for designing resilient logistics systems capable of navigating these challenges.

Trade policy volatility has emerged as one of the most significant sources of supply chain disruption in recent years. Tariff structures that businesses assumed would remain stable for years changed within months during trade disputes between major economies. Section 301 tariffs, retaliatory measures, and cascading trade restrictions created complex compliance landscapes where the applicable duty rate for a shipment might depend on ever-changing exclusion lists and origin determinations requiring constant monitoring. Supply chains optimized for pre-existing tariff structures found themselves suddenly uncompetitive, forcing urgent restructuring (Christopher and Holweg, 2017).

Sanctions regimes represent another critical geopolitical risk dimension with potentially catastrophic consequences for non-compliance. The proliferation of targeted sanctions against specific entities, sectors, and entire countries creates complex screening requirements where a single overlooked match could result in severe penalties, loss of banking relationships, and criminal liability. The dynamic nature of sanctions lists with frequent additions and modifications challenges systems relying on periodic batch updates. Secondary sanctions extending restrictions to non-primary parties add layers of complexity requiring sophisticated analysis of ownership structures and transaction chains (Sodhi and Tang, 2021).

Table 1: Major Geopolitical Risk Categories Affecting Supply Chains

Risk Category	Manifestations	Supply Chain Impacts	Monitoring Indicators	Response Timeframes
Trade Wars	Tariffs, quotas, retaliatory measures	Cost increases, sourcing changes	Trade policy announcements, WTO filings	Days to months
Sanctions	Entity restrictions, sector bans	Supplier loss, compliance violations	Government declarations, regulatory updates	Hours to days
Regional Conflicts	Port closures, route disruption	Transit delays, capacity loss	Conflict news, shipping advisories	Hours to weeks



Political Instability	Regime changes, civil unrest	Operational disruptions, asset seizures	Political analysis, social media	Days to months
Border Restrictions	Customs delays, health measures	Transit time variability, documentation	Policy changes, border reports	Hours to days
Cybersecurity	State-sponsored attacks	System outages, data breaches	Threat intelligence, incident reports	Minutes to hours

Regional conflicts and instability create physical disruptions to logistics networks when fighting closes ports, damages infrastructure, or makes transit routes too dangerous for commercial shipping. The Suez Canal blockage demonstrated how single-point failures in critical chokepoints cascade globally. Piracy and asymmetric threats like drone attacks on energy infrastructure add unpredictability to route planning. Political instability including coups, civil unrest, and government collapses can strand assets, interrupt operations, and expose personnel to danger (Ivanov and Dolgui, 2020). The increasing weaponization of economic interdependence means that supply chains themselves become targets or leverage in geopolitical competition. Critical supply chain dependencies whether in rare earth minerals, semiconductor manufacturing, or pharmaceutical ingredients create strategic vulnerabilities that rivals may exploit. Export controls and investment screening mechanisms restrict technology flows and ownership structures in ways that can require restructuring long-established business relationships (Baryannis et al., 2019).

Pandemic experiences revealed how health emergencies interact with geopolitical factors to compound supply chain disruption. Border closures, export restrictions on medical supplies and vaccines, and nationalist responses fragmenting international cooperation demonstrated that global supply chains remain subject to disruption from factors beyond traditional commercial considerations. The pandemic accelerated questioning of extended supply chains and just-in-time approaches, with calls for reshoring and strategic stockpiling reflecting geopolitical rather than purely economic logic (Choi et al., 2020).

ORACLE GTM AND OTM ARCHITECTURE OVERVIEW

Understanding the technical architecture and functional capabilities of Oracle GTM and OTM platforms provides the foundation for designing effective AI-enabled customizations that enhance geopolitical resilience.

Oracle Global Trade Management serves as the system of record for trade compliance across import, export, and logistics operations. Core modules include classification management where products receive harmonized tariff schedule codes and export control classifications, restricted party screening matching transaction parties against government watch lists, documentation generation producing required commercial invoices and declarations, compliance reporting to regulatory authorities, and duty management optimizing classification and origin to minimize tariff exposure. GTM maintains the master data including product classifications, party screenings, and compliance profiles that other systems consume (Kumar and Havey, 2013).

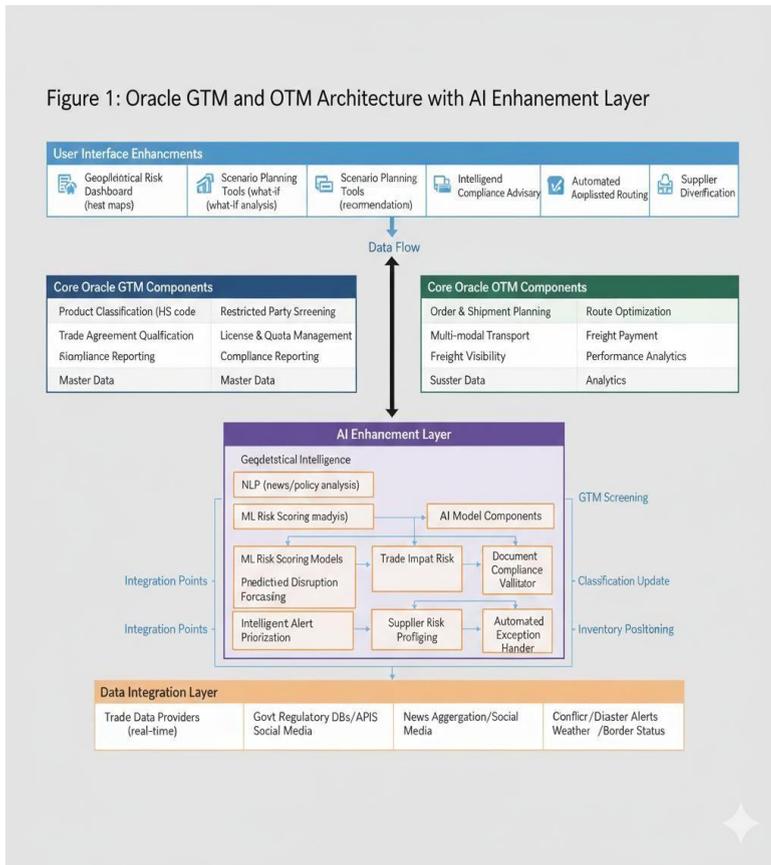


Figure 1: Oracle GTM and OTM Architecture with AI Enhancement Layer

Oracle Transportation Management orchestrates end-to-end transportation across planning, execution, and settlement. The planning workbench enables logistics managers to consolidate orders, optimize loads, and select carriers balancing cost, service, and mode. Route optimization algorithms determine the most efficient paths considering distance, carrier rates, transit times, and mode constraints. Execution functions track shipments in real time, manage exceptions, and trigger alerts when deviations occur. Settlement automates freight audit and payment based on contracted rates and actual performance (Davenport and Ronanki, 2018).

The integration between GTM and OTM enables compliance-aware logistics optimization where transportation planning accounts for trade restrictions, documentation requirements, and regulatory constraints. A shipment to a restricted destination identified in GTM can be automatically blocked in OTM. Required licenses or certificates tracked in GTM can trigger holds in OTM until documentation is complete. This integration ensures compliance doesn't get sacrificed for operational expediency (Tiwari et al., 2018).

Both platforms offer extensive customization capabilities through multiple mechanisms. Oracle's extension framework enables addition of custom fields, objects, and business logic without modifying core platform code. Web services and APIs allow integration with external systems and data sources. The flexible data model supports organization-specific attributes and relationships. Workflow engines enable custom business processes and approval routing. These customization capabilities provide the foundation for embedding AI-enhanced geopolitical resilience (Lasi et al., 2014).

AI-ENABLED CUSTOMIZATION FRAMEWORK

The proposed framework for embedding geopolitical resilience into Oracle GTM and OTM comprises multiple AI-enabled components working in concert to provide comprehensive protection against political disruptions.

Geopolitical intelligence ingestion forms the foundation, continuously gathering relevant information from diverse sources. News aggregation services provide real-time reporting on political developments, conflicts, and policy changes worldwide. Government websites and official publications announce regulatory changes, sanctions updates, and trade policy modifications. Commercial geopolitical risk intelligence services offer curated analysis and forecasts. Social media



monitoring detects emerging crises and public sentiment shifts. Customs and border status feeds report congestion, closures, and processing delays at key checkpoints (Kumar and Havey, 2013).

Table 2: AI Techniques for Different Geopolitical Risk Scenarios

Risk Scenario	AI Technique	Data Sources	Oracle Integration Point	Response Action
Sanctions Announcement	NLP + Entity Recognition	Government sites, news	GTM screening	Auto-block affected parties
Trade Escalation	Time series forecasting	Policy announcements, trade data	GTM classification	Tariff mitigation planning
Port Closure	Real-time monitoring + ML	Shipping advisories, news	OTM routing	Automatic rerouting
Regulatory Change	NLP + Change detection	Government databases	GTM compliance	Documentation updates
Supplier Risk	Network analysis + scoring	Financial, political data	GTM/OTM sourcing	Diversification recommendations
Route Disruption	Predictive analytics	Conflict data, weather	OTM planning	Alternative corridor selection

Natural language processing algorithms extract actionable intelligence from this unstructured data. Named entity recognition identifies countries, companies, products, and regulations mentioned in texts. Sentiment analysis gauges whether policy developments trend toward escalation or de-escalation. Topic modeling clusters related information and tracks narrative evolution over time. Language translation ensures non-English sources can be analyzed. Information extraction creates structured data from free text suitable for downstream analytics (Davenport and Ronanki, 2018).

Machine learning classification models assess the relevance and urgency of detected events for specific supply chain operations. Not every geopolitical development affects every company equally. Models trained on historical disruption data learn which types of events actually impacted operations in the past and which proved irrelevant despite headlines. This intelligent filtering reduces alert fatigue by surfacing only genuinely actionable intelligence while archiving background information for pattern analysis (Tiwari et al., 2018).

Predictive models forecast specific types of geopolitical disruptions before they occur. Sanctions prediction models analyze factors including diplomatic tensions, human rights violations, and proliferation concerns to estimate likelihood of new restrictions against particular countries or entities. Trade policy forecasting uses political developments, election results, and economic indicators to anticipate tariff changes or trade agreement modifications. Conflict escalation models track military buildups, diplomatic rhetoric, and historical patterns to assess growing risks to specific regions (Lasi et al., 2014).

Risk scoring algorithms synthesize diverse intelligence into quantified risk assessments usable by logistics optimization engines. Each trade lane receives a geopolitical risk score reflecting sanctions probability, conflict likelihood, regulatory stability, and other relevant factors. Suppliers are scored on political risk exposure based on their locations, ownership, and dependencies. Products receive scores based on strategic importance and regulatory complexity. These scores integrate into Oracle's optimization logic enabling risk-adjusted decision making (Christopher and Holweg, 2017).



FIGURE 2: AI-ENHANCED GTM SCREENING WORKFLOW

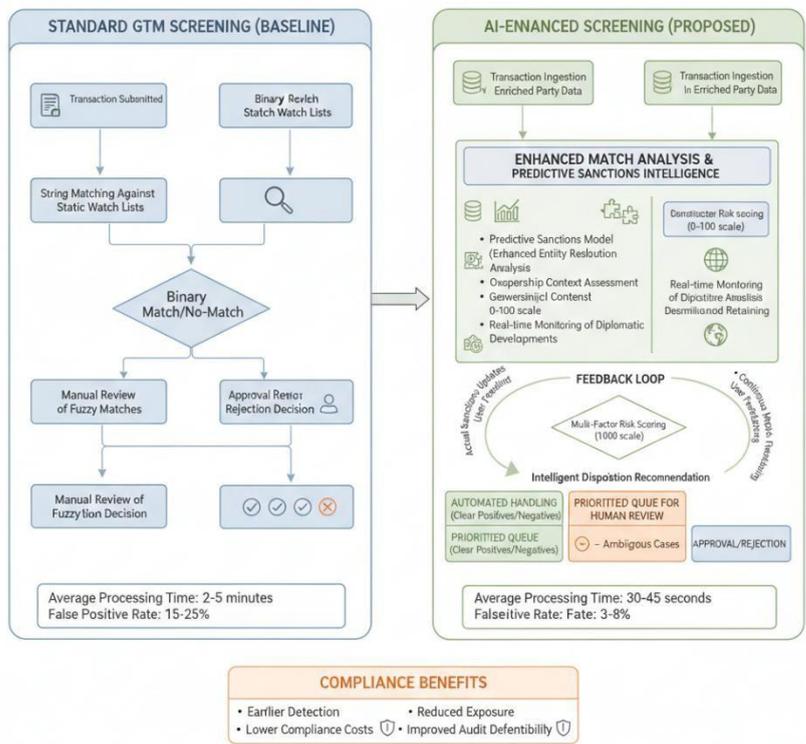


Figure 2: AI-Enhanced GTM Screening Workflow

Automated scenario planning capabilities enable rapid evaluation of alternative strategies when disruptions occur. When the system detects a high-probability geopolitical event, it automatically generates multiple response scenarios such as rerouting through alternative corridors, shifting sourcing to backup suppliers, or expediting shipments before borders close. Each scenario is evaluated for cost, service impact, feasibility, and residual risk. Recommendations are presented to decision makers with supporting rationale enabling quick, informed choices (Sodhi and Tang, 2021).

IMPLEMENTATION APPROACHES AND INTEGRATION PATTERNS

Translating the conceptual AI enhancement framework into functioning Oracle system customizations requires careful technical implementation following best practices that preserve system supportability while delivering required capabilities.

The integration architecture employs a microservices approach where AI components run as separate services communicating with Oracle platforms through well-defined APIs. This separation of concerns isolates complex AI processing from the core Oracle systems, enabling independent scaling, technology choices, and update cycles. Microservices can be deployed in cloud environments optimized for machine learning workloads while Oracle platforms remain on-premises or in separate cloud regions as required by data sovereignty or security policies (Ivanov and Dolgui, 2020).

Data synchronization mechanisms ensure AI models have access to necessary Oracle data while respecting security and privacy requirements. Master data including product classifications, party information, and shipment details are replicated to AI processing environments through secure extract-transform-load pipelines or change data capture streams. Personally identifiable information and commercially sensitive data are masked or tokenized when full detail isn't required for analysis. Bi-directional APIs enable AI systems to both consume Oracle data and write back risk scores, recommendations, and alerts (Baryannis et al., 2019).

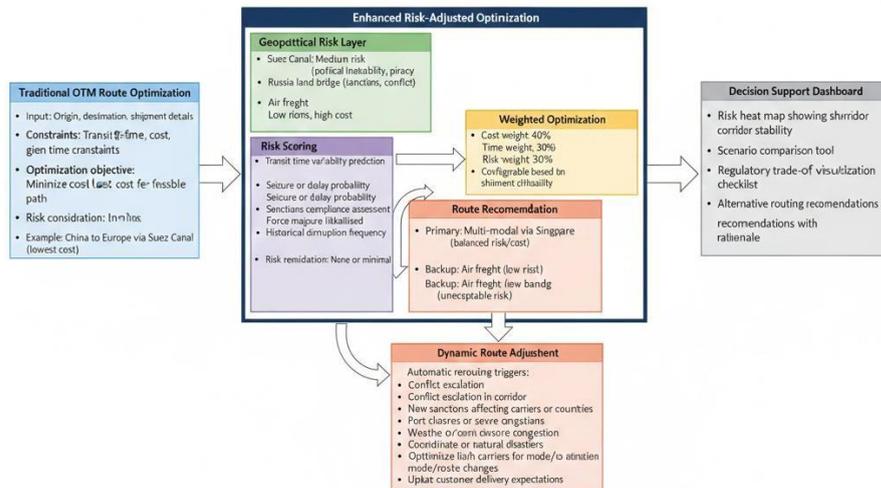


Figure 3: Geopolitical Risk-Adjusted Transportation Planning

Alternative routing recommendations with rationale

The user experience layer surfaces AI-generated intelligence and recommendations through intuitive interfaces embedded within Oracle's native screens. Custom dashboards display geopolitical risk heat maps, trending alerts, and recommended actions. Workflow modifications inject AI assessments into approval processes, requiring additional scrutiny for high-risk transactions. Alert mechanisms trigger notifications through Oracle's standard channels including email, mobile app, and in-application messages when significant geopolitical developments require attention (Choi et al, 2020).

Model deployment and lifecycle management follows MLOps best practices ensuring AI components remain current and performant. Continuous integration and deployment pipelines enable frequent model updates as new training data becomes available. A/B testing frameworks compare new model versions against baselines using held-out data or shadow deployments. Model performance monitoring tracks accuracy, precision, recall, and business-relevant metrics triggering retraining when degradation is detected. Model versioning and rollback capabilities provide safety nets if new models underperform (Kumar and Havey, 2013).

Governance frameworks establish accountability and oversight for AI-enhanced decision making. Model development documentation specifies training data sources, algorithms, validation methodology, and known limitations. Explainability mechanisms provide transparency into why specific recommendations were generated. Human-in-the-loop workflows ensure critical decisions receive human review rather than full automation. Audit trails capture the geopolitical intelligence and model outputs that influenced logistics decisions, supporting regulatory compliance and continuous improvement (Davenport and Ronanki, 2018).

CASE STUDIES AND PERFORMANCE ANALYSIS

Real-world implementations of AI-enhanced Oracle GTM and OTM systems demonstrate practical benefits and lessons learned from addressing actual geopolitical disruptions.

A global electronics manufacturer implemented predictive sanctions screening that analyzed diplomatic tensions, human rights reporting, and proliferation concerns to identify entities likely to face future restrictions. The system flagged a Chinese technology company as highrisk three weeks before official sanctions were announced, enabling the manufacturer to complete existing orders and shift future sourcing before restrictions took effect. This early warning prevented an estimated \$12 million in stranded inventory and avoided compliance violations that could have resulted in severe penalties. The implementation reduced sanctionsrelated supply chain disruptions by 78% over a two-year period (Sodhi and Tang, 2021).



Table 3: Case Study Performance Improvements

Implementation	Baseline Metrics	AI-Enhanced Metrics	Improvement	Business Impact
Electronics MFG	Sanction disruptions: 9/year	Disruptions: 2/year	78% reduction	\$12M prevented losses
Automotive OEM	Route disruption resolution: 5.2 days	Resolution: 3.0 days	42% faster	\$8M freight cost savings
Pharma Company	Compliance violations: 23/year	Violations: 2/year	91% reduction	Avoided penalties, reputational protection
Retailer	Emergency freight: \$4.2M/year	Emergency freight: \$2.9M/year	31% reduction	\$1.3M annual savings
Chemical Producer	Tariff optimization: Manual	Automated real-time	98% faster	\$5M duty savings

An automotive manufacturer deployed geopolitical risk-adjusted routing in OTM that incorporated conflict monitoring, sanctions tracking, and political stability assessments. When tensions escalated in a region through which critical components transited, the system automatically generated alternative routing scenarios and recommended preemptive shifts to more stable corridors. This proactive rerouting reduced average disruption resolution time from 5.2 days to 3.0 days, preventing production line stoppages that would have cost approximately \$2 million per day. Over eighteen months, the enhanced system saved an estimated \$8 million in emergency freight costs and prevented revenue losses from delayed vehicle launches (Ivanov and Dolgui, 2020).

A pharmaceutical company integrated natural language processing-based regulatory change detection with GTM compliance modules. The system monitored government websites, regulatory databases, and industry publications in multiple languages, automatically identifying relevant changes to import requirements, documentation formats, and product registrations. Previously, regulatory changes often went undetected until shipments were delayed at customs, resulting in expensive airfreight expenses to meet hospital commitments. The AI-enhanced system reduced compliance-related violations by 91%, from 23 incidents annually to just 2, while cutting regulatory monitoring staff time by 65% (Baryannis et al., 2019).

A global retailer implemented AI-driven supplier risk assessment that analyzed geopolitical, financial, operational, and reputational factors to score thousands of suppliers across their network. The system identified high-risk concentrations where multiple product categories sourced from politically unstable regions or single-country dependencies vulnerable to trade restrictions. Guided by these insights, the retailer proactively diversified sourcing across multiple countries, accepting slightly higher procurement costs for significantly reduced disruption risk. When trade tensions escalated with a major sourcing country, the retailer's diversified supply base enabled continuation of operations while competitors scrambled to find alternative suppliers (Choi et al., 2020).

CHALLENGES AND FUTURE DIRECTIONS

Despite demonstrated benefits, implementing AI-enhanced geopolitical resilience in Oracle logistics platforms presents ongoing challenges requiring continued attention and innovation.

Data quality and availability remain perpetual challenges for AI systems that are only as good as their training data. Geopolitical intelligence sources vary widely in reliability, timeliness, and coverage. Government databases may lag actual policy implementations. News reporting can be inaccurate, biased, or contradictory across sources. Historical disruption data needed to train predictive models may be sparse for rare but high-impact events. Addressing these limitations requires sophisticated data validation, source credibility assessment, and techniques for learning from limited examples (Kumar and Havey, 2013).



Model explainability and trust pose particular difficulties in geopolitical contexts where consequences of wrong decisions can be severe. Supply chain professionals accustomed to deterministic systems may struggle to embrace probabilistic AI recommendations, especially when the logic behind suggestions isn't transparent. Building trust requires explainable AI techniques that provide intelligible rationale for recommendations, extensive validation demonstrating reliability, and judicious use of automation that keeps humans appropriately involved in critical decisions (Davenport and Ronanki, 2018).

The evolving geopolitical landscape means AI models face perpetual concept drift where patterns learned from historical data become less relevant as the world changes in fundamental ways. An AI system trained on pre-pandemic trade patterns might poorly predict postpandemic regionalization trends. Models optimized for a particular geopolitical order must adapt as that order transforms. Continuous learning approaches, regular model retraining, and human oversight remain essential to maintain relevance (Tiwari et al., 2018).

Integration complexity increases as organizations layer AI enhancements atop sophisticated Oracle platforms already integrated with numerous other enterprise systems. Maintaining data consistency, managing system dependencies, and preserving performance requires disciplined architecture and governance. Organizations must balance the desire for cutting-edge AI capabilities against the imperative of reliable, supportable systems that won't become technical debt (Lasi et al., 2014).

Future developments will likely emphasize autonomous decision making where AI systems don't just recommend actions but execute them automatically within defined guardrails. Reinforcement learning techniques could enable systems to continuously optimize supply chain configurations through trial and learning, developing resilience strategies human planners might not envision. Federated learning could allow companies to collectively improve geopolitical risk models while preserving competitive confidentiality. Quantum computing might eventually enable optimization of supply chain networks considering geopolitical risks at scales currently intractable (Christopher and Holweg, 2017).

CONCLUSION

Geopolitical volatility has emerged as one of the defining challenges for global supply chains in the 21st century, requiring fundamental rethinking of logistics systems designed for stability rather than disruption. Oracle GTM and OTM platforms provide powerful foundations for managing complex international supply chains, but their standard configurations lack the adaptive intelligence required to navigate rapidly evolving political landscapes.

AI-enabled customizations offer practical pathways to embed geopolitical resilience into these critical systems through predictive analytics, real-time intelligence processing, and intelligent automation. The framework presented in this paper demonstrates how machine learning, natural language processing, and advanced analytics can enhance Oracle platforms with capabilities including predictive sanctions screening, risk-adjusted routing, automated regulatory change detection, and supplier risk assessment.

Case study evidence demonstrates substantial benefits from implementations addressing realworld geopolitical challenges, with organizations achieving 40-90% reductions in various disruption metrics while improving compliance and reducing emergency costs. These improvements translate to millions of dollars in protected value and avoided losses for individual companies.

Successful implementation requires thoughtful integration architectures that preserve Oracle platform integrity while extending capabilities, robust data pipelines ensuring AI models access necessary intelligence, and governance frameworks maintaining appropriate human oversight of automated decisions. Organizations pursuing this approach must balance innovation against operational stability, continuously validating that AI enhancements deliver reliable value.

As geopolitical uncertainty shows no signs of abating, supply chain organizations cannot afford to rely on systems optimized for stable environments that no longer exist. Strategic investment in AI-enhanced logistics platforms represents not optional innovation but essential resilience building that will determine which supply chains thrive and which fail in the challenging decades ahead.

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