



The Drivers of Grievance and Unrest in the World's Populations: Understanding Instability, Terrorism & Migration

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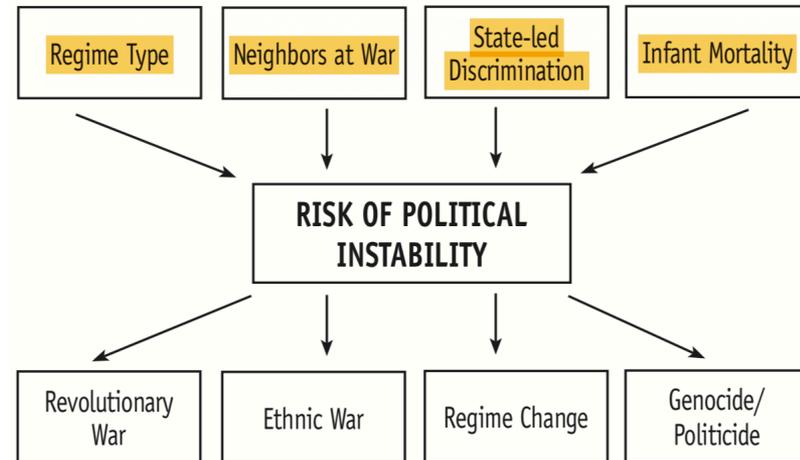
General Concept & Approach

- Great power competition is as much over the hearts and minds of populations as it is about strategic force and control of natural resources.
 - Ligon, Jones & Yager 2019 *The Age of Disruption: How Power Shifts Create More Conflict*. SMA White Paper
- People's grievances and frustrations lead to social unrest, acts of terror and politically destabilizing migration.
- The results of this study anticipate hot spots and their effects on US interests for strategic planning, and identify factors potentially influenced by inter-agency efforts.

Populations are key battlefields in Global Power Competition

- Great powers vie for allegiance of populations
- Great powers manipulate grievances and unrest of populations
- Great power objectives can be frustrated by social unrest
- Products
 - *The Age of Disruption: How Power Shifts Create More Conflict*, Chs. 7,10, 13
 - *Report Aggrieved Populations: Statistical Modeling of Risk and Political Instability in the Influence Environment*
 - *Report Inequality, Risk Sensitivity and Grievance in Context: Summary of Aggrieved Populations Country Reports*
 - Individual reports on 25 countries

- Political Instability Task Force (PITF)
 - 1994 CIA-funded unclassified database of social unrest events, culminated in a series of publications in early 2000s
- Influenced research on political stability, terrorism scales
- Primary finding:
 - Instability predicted 70% of the time with only 4 variables:
 - Weak Democracy
 - Neighboring warfare
 - State-led Discrimination
 - Infant Mortality



Problematics

- Prediction vs theory
- Updating
- New concerns: climate change, food insecurity
- Inequality-driven Risk sensitivity



Aggrieved Populations Project: Concept and Plan

- **Purpose:** Anticipate the Operating Environment 2019-2029 by identifying Emerging Regional and Non-state Challenges

1. Phase I: Global Analysis

- Re-evaluate leading statistical models

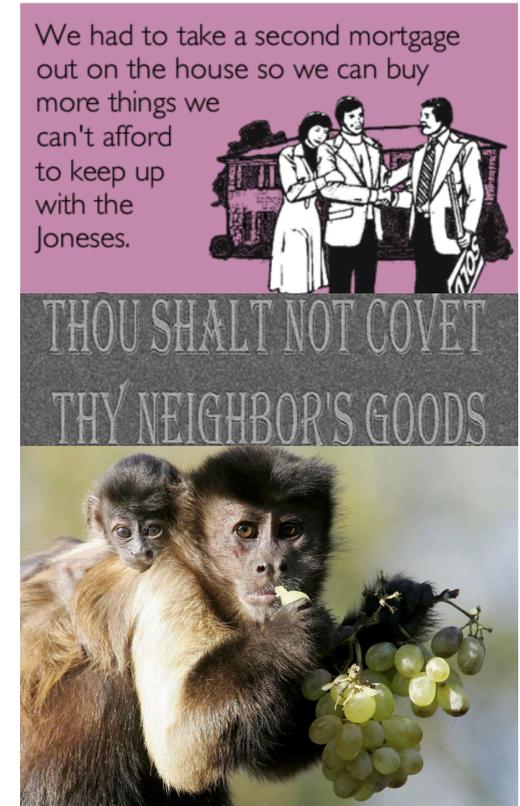
2. Phase II: Country-specific Analysis

- Use risk sensitivity methods to search for social cleavages within 25 key countries
- **Central Concept:** Assess inequality, decision making under risk and political stability

- Control
 - Finland
- Eastern Europe
 - Russia
 - Serbia
 - Croatia
- Western Europe
 - Germany
 - France
 - Great Britain
 - Italy
- North America
 - US
 - Mexico
- Central America
 - Honduras
- South America
 - Brazil
 - Venezuela
- Africa
 - Nigeria
 - Ethiopia
 - South Africa
- East Asia
 - China
 - Indonesia
 - North Korea
 - South Korea
 - Japan
- South Asia
 - India
 - Pakistan
 - Afghanistan
 - Iran

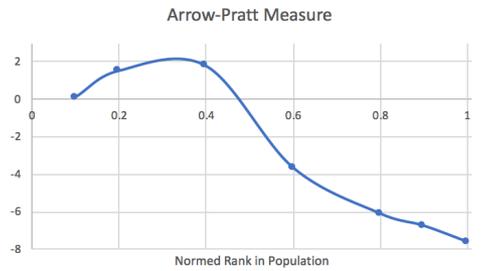
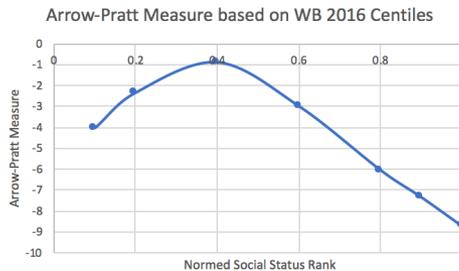
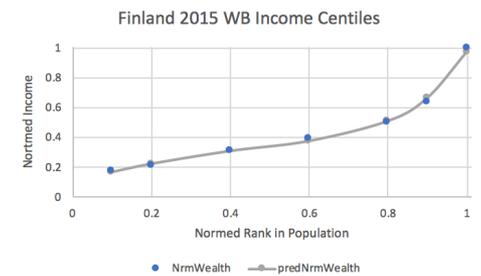
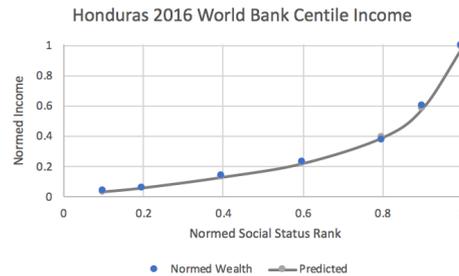
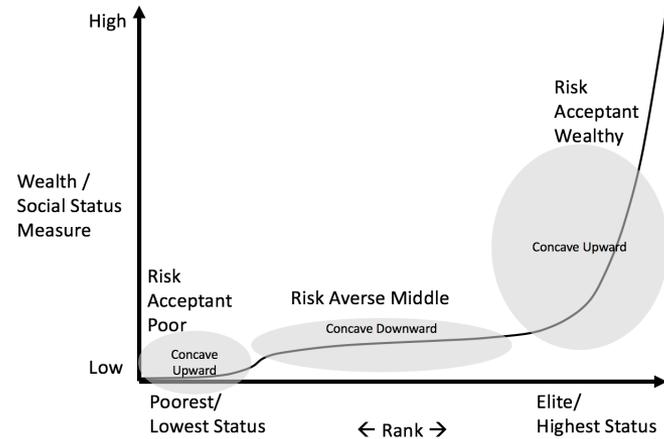
Signaling Status with Wealth

- What if value originated in the social distribution of wealth?
 - Keeping up with the Joneses
 - Deadly Sin of Envy
 - Violating the 10th Commandment
- Friedman, M., & Savage, L. J. (1948). The Utility Analysis of Choices Involving Risk. *Journal of Political Economy*
- People strive to gain tokens of social status (**greed**), resent when they are aware others have more (**envy**), and become distraught when they lose them (**loss aversion**).
 - Even monkeys do it!
 - It's the root of the neuropsychology of fairness and grievance
- Kenneth Arrow & John Pratt propose the Arrow-Pratt measure of risk aversion =
 - $-U(\text{wealth})''/U(\text{wealth})'$



Aggrieved Populations Project: Risk

- Previous research in dozens of societies (tribes, ancient states, communities, modern countries, world) established a pervasive distribution of wealth – expo-sigmoid curve
- Gathered data on percent wealth owned by percentiles of population for 162 countries
- Expo-sigmoid curves fit and used to generate estimates of risk sensitivity



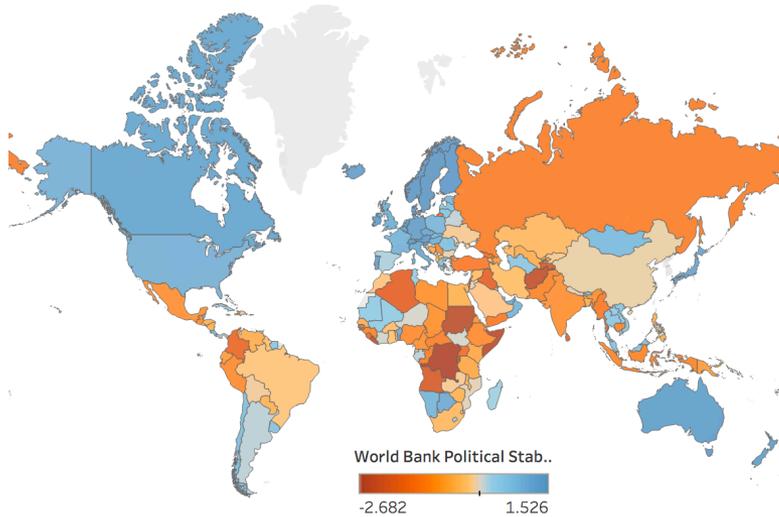
Positive Arrow-Pratt values = Risk Avoidance
 Negative Arrow-Pratt values = Risk Acceptance



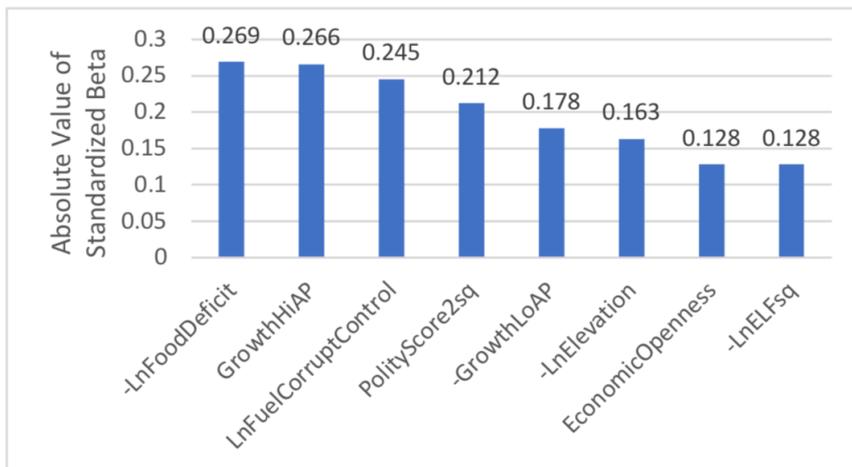
Statistical Approach

- Focused on three dependent variables:
 - **Political Instability** – World Bank Political Stability Scale
 - **Terrorism** – START Terrorism Index
 - **Migration** – UN estimates of net migration
- Began with all variables cited in previous statistical modeling, plus overlooked variables such as:
 - Food insecurity – UN food deficit
 - Impacts from Climate Change – Notre Dame GAIN Index
 - Risk Sensitivity
- Used a stepwise regression and relative value regression to control for multicollinearity and to eliminate variables with no or dubious causality

Political Instability

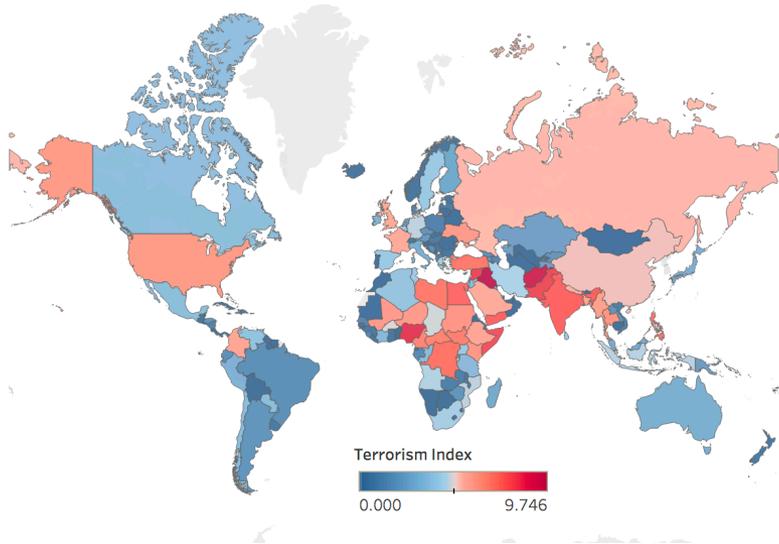


WB Political Stability Index



- **Political Instability Model**
- DV: World Bank Political Stability Scale
- Explanatory Variables:
 - Hunger
 - Risk Acceptant Elites
 - Corruption/Oil Export
 - Weak Democracy
 - Mountainous Terrain
 - Economic Isolation
 - Ethnic division

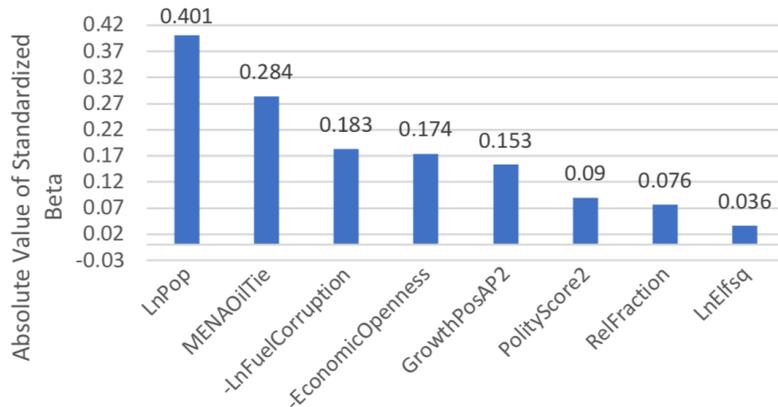
- Hungry people have a grievance, but it takes manipulative and restive elites to mobilize them
- Its not just oil, its corruption +!
- Mountains are difficult to govern
- Engagement with world economic system may create disincentives of elites to defect
- Social divisions are problematic



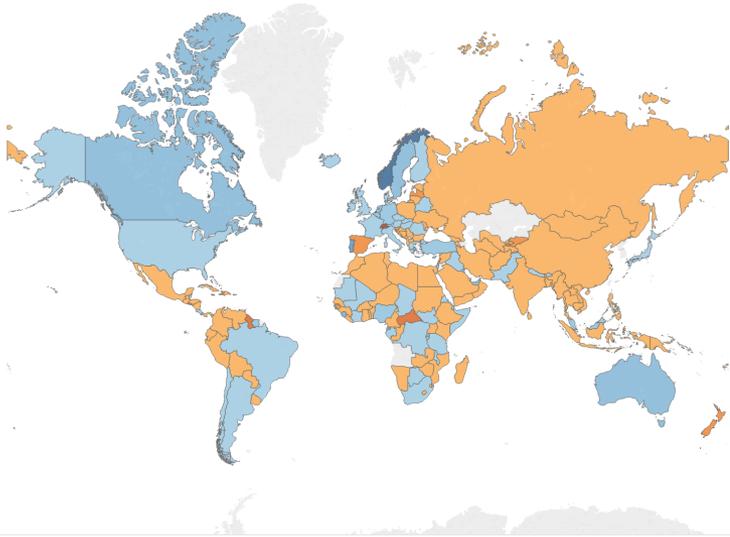
Terrorism Index

- **Terrorism Model**
- DV: Global Terrorism Index
- Explanatory Variables:
 - Large Population
 - Ties to MENA Oil Producers
 - Corruption/Oil Export
 - Status loss among the middle class
 - Weak Democracy
 - Religious division
 - Economic Isolation

- More people = more rare people who will engage in terrorism
- There is something about an oil economy and corruption, and ties to such regimes that is problematic
- Loss aversion creates outrage among middle class – main source of terrorists
- Weak democracies lack capacity to deal with terrorism
- Religion motivates on sacred values



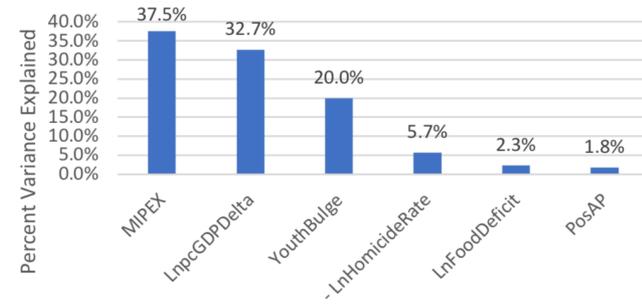
Migration



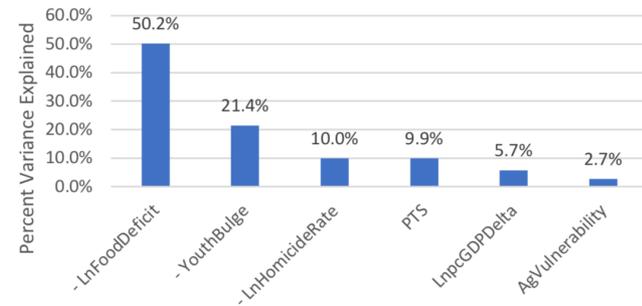
per capita Net Migration
-0.0716 0.1574

Per Capita Net Migration

Immigration to Developed Countries

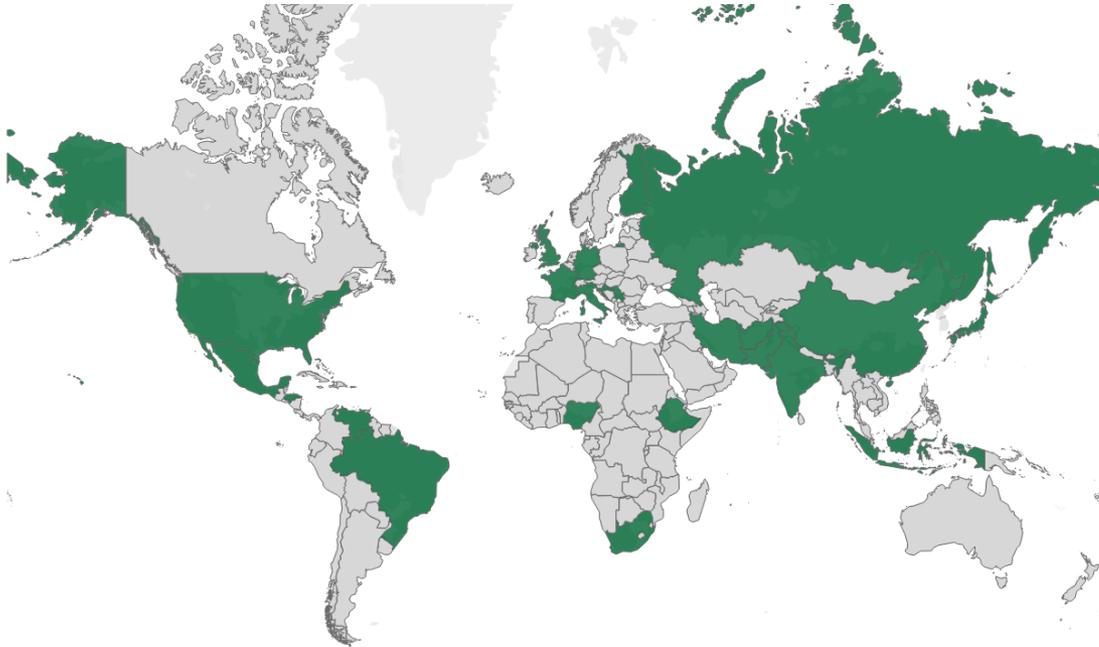


Emigration from Undeveloped Countries



- **Immigration to Developed Countries Model**
- Explanatory Variables:
 - Permissive Immigration Policy (High MIPEX Score)
 - National Wealth (High GDP)
- Wealthy countries with permissive immigration policies attract migrants

- **Emigration from Developing Countries Model**
- Explanatory Variables:
 - Hunger – Food Deficit
 - Youth Bulge
 - Homicide
 - Political terror at home
- People flee hunger, young are able to flee, and people flee homicide & political terror



• Six Dynamics Identified

- **Baselines: Finland and the US**
 - Finland – **low inequality, high stability**
 - US – high inequality + middle class losses from recession
- **Typical Populations**
 - India, Indonesia, Japan, Russia, Serbia and South Korea
- **North Korea**
 - Masses just try to survive, intense intrigue & competition among elite; Kim family uses terror to contain dissent

- **Risk Acceptant Populations**
 - Brazil, Honduras, Mexico, Nigeria, South Africa, and Venezuela
 - Primary manifestation – **Homicide & Emigration**
- **Loss Averse Populations**
 - Europe, Pakistan, Venezuela, Iran
 - Angry middle class, protest, political shifts
 - Sanctions exacerbate these effects in Iran
 - Primary manifestation - **Nationalism**
- **Agrarian Populations**
 - Afghanistan, Ethiopia, Honduras, India, Indonesia, Nigeria and Pakistan
 - High inequality and **competition/unrest in rural areas**



Summary Findings

Global country-level study:

- **Political instability** is driven by hunger, risk acceptant elites, the interaction of fuel export and corruption, weak democracy, mountainous terrain, economic isolation, and ethnic division.
- **Terrorism** is fueled by large populations, ties to MENA oil producers, the interaction of fuel export and corruption, economic isolation, and a risk acceptant middle class.
- **Migration from undeveloped countries** is driven by hunger, a youth bulge, homicide and political oppression, and
- **Migration to developed countries** is driven by permissive immigration policies and the attraction of national wealth.

Country-specific statistical analyses revealed several patterns of stability and instability based on the inequality and risk sensitivity of their populations.

- Countries with **low inequality** are **stable**, such as Finland.
- Countries with **unusually high levels of inequality** are characterized by extremely **high levels of interpersonal violence**, such as Honduras and South Africa. Interpersonal violence is a driver of **illegal migration**.
- **Agrarian countries** have extremely **high levels of inequality** and consequently **experience unrest** in rural areas, which in turn is exacerbated by rural/urban inequalities. Afghanistan, Pakistan, Honduras, Nigeria and Ethiopia are good examples.
- Countries where some **sectors have lost wealth and status**, or perceive an external threat to their status, have **seen nationalist and populist parties gain power**. Examples include Pakistan, Germany, Italy, UK, and to a lesser extent Iran.