



Global Attractive Index for Money laundering using Walker's Gravity Model: An Approach to Determine the Preferred Destinations

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Abstract

This study develops a Money Laundering Attractiveness Index (MLAI) for the countries that are preferred for money laundering. The study first discussed the concept of money laundering and the size and volume of money laundered. Walker's Gravity Model has been used to determine the preferred destinations. The model was developed by John Walker to estimate the illegal funds flow globally. The model consists of two segments, including attractiveness and distance. Data from 2002 to 2022 has been collected and analyzed across approximately 150 countries. The study examined the flow of illegal funds from the source country to the preferred destination among all income groups classified by the World Bank. The results show that countries with stable economic conditions and a higher standard of living hold the top positions in the money laundering attractiveness index. The study underscores the need for coordinated global action and recommends revisiting current anti-money laundering strategies. The proposed index offers a valuable tool for intergovernmental bodies and policymakers to develop more targeted and effective responses against organized crimes like money laundering.

Keywords: Illegal Funds Flow, Money Laundering, Money Laundering Attractiveness Index, Organized Crime, Walker's Gravity Model, World Bank



Introduction

Money laundering is a global financial crime that helps wrongdoers secure their illegal wealth, worth trillions of dollars. It is a crime that makes other criminal activities fruitful, as it aids in utilizing such funds without getting caught.

Money laundering is a crime traced back to 13 Century BC.¹ It has become so sophisticated and developed that it helps the criminal elements to wash their dirty money, amounting to trillions of dollars generated through corruption, public fraud, kickbacks, embezzlement of funds, trafficking of arms, drugs or humans, smuggling of goods, extortion, and other organized crimes. According to the United Nations Office on Drugs and Crime estimated amount of money laundered each year is 2% to 5%², if calculated on the higher side, it will be more than USD\$ 5 trillion. The quantum of the amount involved in the crime of money laundering solely explains the need to address the issue.

The dynamic and ambiguous nature of the money laundering process, along with the significant amount involved, makes it challenging to estimate the total funds involved in money laundering. Gerbrands et al (2022) also suggest the same. The concept of money laundering is not new, and for centuries the people and businesses have been using such techniques. The term money laundering came into existence when mafias in the United States started to use public laundry businesses to convert their illegal funds due to cash cash-based nature of their business (Korejo, 2021). The laws and regulations related to money laundering have been available for decades. However, it was after the 9/11 terrorist attacks on American soil that revived the anti-money laundering regimes.

There is also a lack of a single, mutually agreed-upon definition of money laundering. Different intergovernmental bodies define money laundering differently, which creates discrepancies in laws and regulations implemented in various jurisdictions. However, in general, it is mutually agreed that money laundering is the set of complex transactions that consists of three stages. Placement, at this stage, illicit funds are introduced into the legitimate financial system. In the second stage of Layering, the origin of the money is concealed through complex transactions to obscure its source. The third and final stage of Integration is when laundered money is reintroduced into the economy as apparently legitimate assets or funds. Below are some of the available definitions:

United Nations

“The conversion or transfer of property, knowing that such property is derived from any offense(s), for the purpose of concealing or disguising the illicit origin of the property or of assisting any person who is involved in such offense(s) to evade the legal consequences of his actions”.

FATF

¹ IBA Money Laundering Forum

² The IMF and World Bank estimate that 3%-5% of global GDP (<https://2009-2017.state.gov/j/inl/c/crime/c44634.htm>)



"Money laundering" is the act by which the proceeds of crime are made to appear legitimate.

EU

Money laundering is the process by which criminals conceal the illegal origin of their property or income.

Anti-money laundering (AML) regulations enforced around the world generally focus on countries with weak governance, endemic corruption, and strong criminal ties, usually less developed or developing nations. These countries are subjected to a large resource drain resulting from illegal financial outflows with negative consequences for poverty reduction, education, health, infrastructure and economic growth. When the UN office on drugs and crime published an analysis of 11,000 cases of money laundering in 2019, the data showed that the vast majority of the money gets sent to offshore jurisdictions where, as George Staple, a retired spy, points out, it cannot be seized (Mugarura, 2020; Sharman, 2020). Early theories posited that much of the laundered money would have gone back to the country to fund other criminal activities, but the laundering trail suggests that criminals feel more secure if they keep their wealth offshore and are thus insulated from getting shut down by authorities who go after fleeced victims. This is a fact that undermines the current world approach to anti-money laundering (AML), because that approach is more concerned with the source and first port of call rather than the journey to shelter for stolen money. USD 3.7 billion in corruption-related assets were identified in richer countries by Transparency International to illustrate this problem. Christensen (2011) claims that a political unwillingness is creating barriers to action on implementation. As such, reliable estimates of the flows of money through the system and preferred laundering sectors become paramount, where Walker's Gravity Model becomes an invaluable framework in the analysis of movement and risk.

Money laundering presents significant economic and social threats globally, as it aids criminal activity and denies wealth to nations that is rightfully theirs. Although researchers have focused more on the concept of money laundering, its definition, the effects and other theoretical aspects of the phenomenon (e.g., Quirk, 1997; Walker, 2002). A review of several studies highlights that there is still a large gap concerning the attractive destinations where dirty money is laundered. For example, most studies only analyze single countries, i.e., the Netherlands (Unger et al., 2006), Romania-Turkey (Ene, 2014), or Russia or Indonesia (Romana et al., 2023; Ariesiyani & Alham, 2023), providing incomplete answers. The Basel AML Index is a good effort so far, but it focuses mainly on source jurisdictions, not destination countries. There are no global rankings that point out the best or worst places to stash laundered money. This paper seeks to address the gap by being the first to create a worldwide index of desirable places to launder money.

Money laundering causes widespread economic, social, and political harm by fueling crime, distorting markets, weakening institutions, and reducing public resources. It undermines democracy, increases inequality, and damages international reputation, leading to sanctions and reduced global cooperation. Ultimately, it threatens national security and long-term development stability.



Review of Related Literature

The previous literature in this section addresses multiple aspects of money laundering, measures and its socio-economic effects. It offers a window into techniques employed to quantify proceeds of crime and assesses money laundering impacts on development, stability and governance.

Unger et al. (2006), by using Walker's Gravity Model, stated that some USD 30 billion is laundered annually in the Netherlands. The research also underlined the importance of accurate information to strengthen AML frameworks. Schneider (2010) also employed the MIMIC model, and estimated that the turnover in organized crime in 20 OECD countries amounted to a considerable USD 2.013 trillion. These analyses emphasize the difficulties around data availability and the changing styles of laundering methods.

Gjoni et al. (2015) linked globalization and technological innovation to rising money laundering, particularly affecting nations with weak AML laws. Mugarura (2020) echoed that less-developed countries bear an unfair AML burden. Korejo et al. (2021) criticized the evolving FATF framework for weak political commitment. Shah and Aish (2021) tied laundering to macroeconomic instability in South Asia. Gilmour (2022) emphasized evolving laundering methods via digital tools. Studies by UNODC (2011), Ene (2014), Ardizzi et al. (2014), Chong and Silanes (2015), Alouk and Bagheri (2012), Otusanya (2011), Ali (2022), and Harris et al. (2009) explored laundering's institutional, developmental, and economic consequences.

Recent studies on money laundering show a strong shift toward using advanced technologies like artificial intelligence to detect and prevent illegal financial activities. For example, Deprez et al. (2024) found that graph-based models, especially graph neural networks, are much better at uncovering suspicious connections between transactions than traditional methods. In a follow-up study, they emphasized how important it is for these systems to keep learning and adapting over time to stay ahead of new laundering tactics (Deprez et al., 2025). Likewise, Fan et al. (2025) developed a deep learning framework specifically for mobile money transactions, helping systems detect unusual patterns while also protecting user privacy. Jensen and Iosifidis (2022) added that while machine learning offers powerful tools, it's just as important to ensure these systems are transparent, fair, and built on reliable data. Together, these efforts point to a future where smart, adaptable technologies play a central role in fighting financial crime.

Alongside these tech-focused approaches, researchers are also exploring the broader economic and political impacts of money laundering. Khelil et al. (2024) show how laundering contributes to serious problems like rising inflation, tax avoidance, and weaker financial accountability. In Malta, Mintoff and Vella (2024) reveal how laundering networks often operate with political protection, making it even harder to enforce the law. On a more global scale, Ahuja et al. (2023) trace how academic interest in AML has grown over the years, especially after 2012, as concerns around cryptocurrency and digital finance expanded. Al Naqbi et al. (2025) confirm this trend, highlighting the need for blockchain-based tools and AI to respond to the growing risks of crypto-related financial crime. Altogether, these studies reflect how the fight against money laundering



requires not just better technology, but also stronger institutions and smarter, more targeted policies.

In general, the literature highlights financial crime as being a key factor undermining national economies, governance and development, with a need for concerted international cooperation, accurate information and flexible policy responses to address the fast-evolving dimensions of this growing problem.

Methodology and Data

The previous sections highlight the need for quantification of the amount involved in money laundering to counter such activities. In order to implement more effective and fruitful laws and regulations policy policymakers require the exact areas and jurisdictions to focus on. Therefore, to determine the most favorite destinations for laundering of ill-gotten wealth, a suitable approach is required, the Walker Gravity Model best fits the said purpose.

Sample Size

Data has been collected across approximately 150 countries to determine of most favorable destinations for money laundering through Walker’s Gravity Model, due to the unavailability of data number of countries varies from year to year. The list of countries is mentioned in Table 1. Data collection to establish the Money Laundering Attractiveness Index involves various observations and stats from year 2002 to 2022 for each country and includes economic and financial data like GDP per capita, financial deposit (FD). Perception indicators consist of government attitude (GA), banking secrecy (BS), conflict (CF), and corruption (C). SWIFT and the Egmont group are also taking part. The variables are briefly discussed in Table 2.

The second part of the study aims to determine the preferred destination for money laundering purposes from countries bifurcated in income classification based on GNI per capita by the World Bank. The said list comprises of 218 countries distributed as high-income countries (> 13,845), upper middle countries (4,466 - 13,845), lower middle countries (1,136 - 4,465), and low-income countries (<= 1,135).

Table 1
 List of 154 total countries used for Estimations

Afghanistan	Cyprus	Kuwait	Russia
Albania	Czech Republic	Kyrgyzstan	Samoa
Algeria	Denmark	Laos	Saudi Arabia
Argentina	Dominica	Latvia	Senegal
Armenia	Dominican Republic	Lebanon	Serbia
Australia	East Timor	Lesotho	Seychelles
Austria	Ecuador	Liberia	Sierra Leone
Azerbaijan	Egypt	Libya	Singapore
Bahamas, The	El Salvador	Lithuania	Slovakia
Bahrain	Fiji	Luxembourg	Slovenia
Bangladesh	Finland	Macau	Solomon Islands



Belarus	France	Macedonia	South Africa
Belgium	Gambia, The	Madagascar	Spain
Belize	Georgia	Malawi	Sri Lanka
Benin	Germany	Malaysia	Sudan
Bhutan	Ghana	Maldives	Suriname
Bolivia	Greece	Mali	Swaziland
Bosnia & Herzegovina	Grenada	Malta	Sweden
Botswana	Guinea	Mauritania	Switzerland
Brazil	Guinea-Bissau	Mauritius	Tajikistan
Brunei	Guyana	Mexico	Tanzania
Bulgaria	Haiti	Moldova	Thailand
Burkina Faso	Honduras	Mongolia	Togo
Burundi	Hong Kong	Nepal	Trinidad & Tobago
Cambodia	Hungary	Netherlands	Tunisia
Cameroon	Iceland	Nicaragua	Turkey
Canada	India	Niger	Uganda
Cape Verde	Indonesia	Nigeria	Ukraine
Central African Rep.	Iran	Oman	United Arab Emirates
Chad	Iraq	Pakistan	United Kingdom
Chile	Ireland	Panama	United States
China	Israel	Papua New Guinea	Uruguay
Colombia	Italy	Paraguay	Vanuatu
Comoros	Jamaica	Peru	Venezuela
Congo, Dem. Rep.	Japan	Philippines	Vietnam
Congo, Repub. of the	Jordan	Poland	Yemen
Costa Rica	Kazakhstan	Portugal	Zambia
Cote d'Ivoire	Kenya	Qatar	
Croatia	Korea, South	Romania	

Walker's Gravity Model

Walker's Gravity Model originated from Newton's Gravity model, proposed by Isaac Newton in 1687, same was later transformed by Nobel Prize winner Jan Tinbergen for bilateral trade estimations. Unlike other techniques, Walker's formula does not only rely on official statistics, thus, estimates are not based only on published or officially provided data, which, in the case of money laundering and organized crime is built on estimates and detected cases, and never shows the actual picture. This formula has better country coverage and only considers the amount when it enters in a country for the purpose of laundering.

$$P(X, y_i) = \frac{1}{\sum_{i=1}^n \left[\frac{\text{ATTRACTIVENESS}(y_i)}{\text{dist}(X, y_i)} \right]} \times \frac{\text{attractiveness}(y_i)}{\text{dist}(X, y_i)}$$

Where P is the proportion of money flowing from country X to country y_i

In our case, X is a specific country, y_i that is another country i



$i=1, \dots, n$ for all countries of the world.

To apply the Gravity Model, attractiveness has to be calculated for all destination countries relative to the source country. Only then you can calculate how the money from the source country X is distributed all over the world (dividing by the sum of the attractiveness of all countries makes sure that the total is 100%). (Khan et al. 2018)

Walker’s Gravity Model further disseminates as follows:

$$\text{Attractiveness} = \text{GPD} * (3 * \text{BS} + \text{GA} + \text{SWIFT} + \text{FD} - 3 * \text{CF} - \text{CR} - \text{EG} + 10)$$

$$\text{Distance} = \text{language} + \text{trade} + \text{colonial background} + \text{physical distance}$$

Table 2
 Gravity Model Variables – Definition, Justification & Data Source³

Variable	Definition & Justification	Data Source & Link
GDP per capita	Reflects economic development and laundering opportunities. Higher GDP attracts more illicit funds due to complex financial systems.	World Bank Data Bank
Banking Secrecy	Scale 1–4 based on legal secrecy. Strong secrecy jurisdictions attract launderers seeking anonymity.	Financial Secrecy Index
Government Attitude	Score 0–4 based on FATF membership. Indicates AML commitment; lax enforcement signals tolerance toward laundering.	FATF
SWIFT Membership	0 = non-member, 1 = member. Shows global financial integration; higher connectivity may facilitate illicit transfers.	SWIFT
Financial Deposits	Deposit-to-GDP ratio; higher figures suggest deeper financial systems that can be misused for laundering.	World Bank Data Bank
Conflict	Score 0–4 based on current/past conflict status. Unstable nations may deter or attract laundering, depending on risk.	UCDP Conflict Data
Corruption	Scale 1–5; high corruption may attract collusive laundering but increases uncertainty.	Transparency International – CPI
Egmont Group	0 = not a member, 1 = member. Membership indicates strong AML institutions and cooperation, reducing laundering risk.	Egmont Group
Language	0 = same, 1 = different. Shared language eases illicit financial coordination.	CIA World Factbook - Languages

³ For full details of variables, refer to Unger et al. (2006) and Khan et al. (2018)



Colonial Background	0 = same, 1 = different. Shared history fosters legal and institutional similarities, enabling illicit ties.	CIA World Factbook - Background
Trade	0 = similar partners, 1 = different. Closer trade ties may mask laundering via misinvoicing.	Exports / Imports
Distance	Real capital-to-capital distance in km; proximity encourages flows (licit or illicit).	DistanceFromTo

Data Analysis and Results

With the aim of establishing an index for the most preferred destination for money laundering, data have been collected with respect to 139 countries. The period covered is from year 2002 to 2022, and data have been collected for more than 150 countries. The enclosed Tables 03 to 06 show the top twenty.⁴ The most preferred jurisdictions for the purpose of money laundering, whereas Table # 03 shows the most common countries in the entire index. The count column reflects the number of times a country appeared in a year in the attractiveness table of any country, and the sum is the total attractiveness score of a country in a year. Further, the score is averaged out by dividing it by the total count.

⁴ Due to publication space limitations, only the top 20 countries are listed. The complete index is available upon request from the author and may be used for future research with proper acknowledgment.

Table 3
Money Laundering Attractiveness Index

2022				2021				2020				2019				2018			
Country	Count	Sum	Average	Country	Count	Sum	Average	Country	Count	Sum	Average	Country	Count	Sum	Average	Country	Count	Sum	Average
Luxembourg	135	2,551.24	18.90	Switzerland	138	820.24	5.94	Switzerland	138	845.52	6.13	Switzerland	138	878.21	6.36	Switzerland	138	958.52	6.95
Switzerland	135	797.30	5.91	Luxembourg	137	2,808.77	20.50	Luxembourg	137	2,701.42	19.72	Luxembourg	137	2,668.63	19.48	Luxembourg	137	899.43	6.57
Hong Kong	135	702.42	5.20	Hong Kong	135	731.71	5.42	Hong Kong	136	777.09	5.71	Hong Kong	136	773.23	5.69	Sweden	137	488.74	3.57
Qatar	135	471.45	3.49	Netherlands	135	338.83	2.51	Austria	135	348.82	2.58	Austria	135	338.12	2.50	Denmark	137	682.62	4.98
Austria	135	321.69	2.38	Austria	135	336.52	2.49	Netherlands	135	334.54	2.48	Netherlands	135	325.39	2.41	Netherlands	135	442.04	3.27
Ireland	135	317.86	2.35	Ireland	135	329.45	2.44	Ireland	135	320.24	2.37	Ireland	135	296.96	2.20	Austria	134	379.90	2.84
Netherlands	135	316.67	2.35	Denmark	135	237.14	1.76	Denmark	135	245.49	1.82	Denmark	135	238.29	1.77	Iceland	134	246.50	1.84
Sweden	134	240.41	1.79	United Kingdom	134	312.07	2.33	Sweden	134	257.32	1.92	United Kingdom	134	338.14	2.52	Finland	133	288.63	2.17
United Kingdom	133	297.80	2.24	Sweden	134	269.18	2.01	United Kingdom	133	299.40	2.25	Sweden	134	233.64	1.74	France	133	322.43	2.42
Germany	130	254.58	1.96	Germany	131	275.22	2.10	Qatar	132	393.82	2.98	Qatar	133	409.26	3.08	United Kingdom	132	397.38	3.01
Iceland	129	198.78	1.54	Finland	127	191.79	1.51	Germany	132	287.46	2.18	Germany	133	280.80	2.11	Qatar	128	398.11	3.11
Israel	128	313.88	2.45	Iceland	127	188.23	1.48	Finland	131	203.54	1.55	Iceland	130	205.13	1.58	Cyprus	126	341.81	2.71
Denmark	127	218.11	1.72	Qatar	126	373.93	2.97	France	129	244.49	1.90	Cyprus	124	285.85	2.31	Germany	125	298.91	2.39
Belgium	122	309.82	2.54	France	126	234.09	1.86	Belgium	122	340.14	2.79	Belgium	122	334.59	2.74	Hong Kong	115	442.30	3.85
France	117	204.71	1.75	Israel	124	298.81	2.41	Japan	116	302.22	2.61	France	118	210.38	1.78	United States	115	376.82	3.28
Singapore	115	490.69	4.27	Belgium	122	330.70	2.71	Cyprus	114	248.16	2.18	Finland	116	171.43	1.48	Japan	110	282.93	2.57
Cyprus	111	240.55	2.17	Cyprus	114	256.62	2.25	Israel	112	258.96	2.31	Japan	115	291.16	2.53	Belgium	103	232.57	2.26
Finland	110	158.63	1.44	Singapore	113	454.73	4.02	Iceland	110	161.34	1.47	Singapore	108	374.51	3.47	Singapore	97	354.97	3.66
United Arab Emirates	101	271.76	2.69	Japan	110	271.08	2.46	Singapore	108	392.46	3.63	Israel	91	202.76	2.23	Canada	90	208.24	2.31
Japan	88	201.23	2.29	Italy	95	178.12	1.87	Italy	93	172.52	1.86	Italy	83	153.79	1.85	Macau	87	353.64	4.06

Table 4
Money Laundering Attractiveness Index

2017				2016				2015				2014							
Country	Count	Sum	Average	Country	Count	Sum	Average	Country	Count	Sum	Average	Country	Count	Sum	Average	Country	Count	Sum	Average
Switzerland	138	934.27	6.77	Switzerland	138	934.27	6.77	Switzerland	138	935.61	6.78	Switzerland	138	922.129422	6.68209729	Switzerland	138	922.129422	6.68209729
Luxembourg	137	862.49	6.30	Luxembourg	137	862.49	6.30	Luxembourg	137	846.58	6.18	Cyprus	137	547.431265	3.99584865	Luxembourg	137	547.431265	3.99584865
Sweden	137	480.18	3.50	Sweden	137	480.18	3.50	Sweden	137	470.79	3.44	Denmark	137	692.511114	5.05482565	Denmark	137	692.511114	5.05482565
Denmark	137	670.31	4.89	Denmark	137	670.31	4.89	Denmark	137	680.56	4.97	Sweden	137	464.274113	3.38886214	Sweden	137	464.274113	3.38886214
Netherlands	135	447.87	3.32	Netherlands	135	447.87	3.32	Qatar	135	447.63	3.32	Luxembourg	136	737.84	5.43	Cyprus	136	737.84	5.43
Cyprus	134	456.03	3.40	Cyprus	134	456.03	3.40	Cyprus	135	493.05	3.65	Netherlands	135	453.09	3.36	Netherlands	135	453.09	3.36
Austria	134	362.32	2.70	Austria	134	362.32	2.70	Netherlands	135	454.00	3.36	Finland	135	277.12	2.05	Ireland	135	277.12	2.05

Finland	134	283.55	2.70	Finland	134	283.55	2.12	Austria	134	364.22	2.72	Austria	134	381.71	2.85	Austria
United Kingdom	132	389.26	2.95	United Kingdom	132	389.26	2.95	Finland	134	279.57	2.09	Greece	134	447.98	3.34	Finland
France	132	305.48	2.31	France	132	305.48	2.31	United Kingdom	132	384.06	2.91	Ireland	134	284.31	2.12	Iceland
Qatar	131	426.31	2.25	Qatar	131	426.31	3.25	France	132	291.24	2.21	United Kingdom	132	388.36	2.94	United Kingdom
Germany	127	292.34	2.30	Germany	127	292.34	2.30	Germany	127	287.36	2.26	Germany	125	286.30	2.29	France
Iceland	125	219.40	1.76	Iceland	125	219.40	1.76	Iceland	117	200.64	1.71	France	125	275.96	2.21	Germany
United States	119	390.26	3.28	United States	119	390.26	3.28	Hong Kong	112	411.44	3.67	Hong Kong	118	440.23	3.73	Spain
Hong Kong	115	433.74	3.77	Hong Kong	115	433.74	3.77	United States	112	366.91	3.28	Iceland	115	200.20	1.74	Hong Kong
Japan	108	276.81	2.56	Japan	108	276.81	2.56	Ireland	103	198.96	1.93	United States	114	369.04	3.24	United Kingdom
Ireland	106	208.35	1.97	Ireland	106	208.35	1.97	Kuwait	102	249.85	2.45	Japan	107	267.67	2.50	Japan
Singapore	91	339.40	3.73	Singapore	91	339.40	3.73	Japan	101	257.59	2.55	Singapore	101	354.07	3.51	Italy
Kuwait	86	202.18	2.35	Kuwait	86	202.18	2.35	Singapore	91	334.76	3.68	Spain	99	195.52	1.97	Singapore
Spain	85	155.15	1.83	Spain	85	155.15	1.83	Spain	90	167.69	1.86	Qatar	95	280.89	2.96	Qatar

Table 5
Money Laundering Attractiveness Index

2012				2011				2010				2009				2008			
Country	Count	Sum	Average	Country	Count	Sum	Average	Country	Count	Sum	Average	Country	Count	Sum	Average	Country	Count	Sum	Average
Switzerland	138	923.40	6.69	Switzerland	137	868.72	6.34	Switzerland	137	859.70	6.28	Switzerland	137	853.29	6.23	Switzerland	173	1,078.44	6.23
Luxembourg	137	751.22	5.48	Cyprus	136	606.30	4.46	Luxembourg	136	774.91	5.70	Luxembourg	136	866.26	6.37	Luxembourg	169	1,128.83	6.68
Cyprus	137	596.36	4.35	Luxembourg	136	741.36	5.45	Cyprus	136	590.04	4.34	Cyprus	136	582.84	4.29	Denmark	151	896.78	5.94
Denmark	137	737.72	5.38	Denmark	136	732.87	5.39	Denmark	136	758.38	5.58	Denmark	136	782.67	5.75	Cyprus	136	604.01	4.44
Sweden	137	467.53	3.41	Sweden	136	449.28	3.30	Sweden	136	437.66	3.22	United Kingdom	136	526.51	3.87	Ireland	136	525.26	3.86
Netherlands	136	469.21	3.45	Netherlands	136	456.53	3.36	Ireland	136	415.10	3.05	Ireland	136	506.55	3.72	Finland	135	277.27	2.05
Ireland	136	357.61	2.63	Austria	135	407.93	3.02	Netherlands	136	450.82	3.31	Sweden	136	424.31	3.12	Sweden	132	417.33	3.16
United Kingdom	135	455.09	3.37	United Kingdom	135	466.37	3.45	Iceland	136	352.90	2.59	Netherlands	136	461.91	3.40	France	130	276.03	2.12
Finland	135	289.10	2.14	Ireland	135	362.85	2.69	United Kingdom	135	502.49	3.72	Iceland	136	390.49	2.87	Iceland	129	446.29	3.46
Iceland	135	275.99	2.04	Iceland	135	306.23	2.27	Finland	134	274.59	2.05	Austria	135	416.45	3.08	United Kingdom	128	546.47	4.27
Austria	134	421.75	3.15	Finland	134	281.12	2.10	Spain	132	319.17	2.42	Finland	135	265.29	1.97	Netherlands	128	433.89	3.39
Germany	130	308.88	2.38	Spain	131	304.10	2.32	France	129	283.35	2.20	Spain	132	324.37	2.46	Spain	128	320.11	2.50
Spain	129	282.95	2.19	France	128	288.71	2.26	Germany	128	297.21	2.32	Germany	129	319.08	2.47	Germany	128	326.64	2.55
Italy	122	260.40	2.13	Germany	127	303.95	2.39	Austria	123	480.43	3.91	France	128	277.57	2.17	Italy	128	250.07	1.95
Hong Kong	112	392.28	3.50	Italy	120	261.69	2.18	Italy	118	255.48	2.17	Italy	118	243.72	2.07	Japan	120	315.27	2.63
United States	109	350.15	3.21	Hong Kong	108	385.17	3.57	Greece	112	228.45	2.04	Qatar	114	327.34	2.87	United States	116	342.78	2.95
Japan	105	268.32	2.56	United States	103	334.01	3.24	United States	108	363.13	3.36	United States	114	367.88	3.23	Greece	112	203.18	1.81
Canada	91	211.37	2.32	Japan	100	252.74	2.53	Japan	106	269.31	2.54	Kuwait	110	253.27	2.30	Portugal	110	198.26	1.80
Qatar	88	256.62	2.92	Greece	86	185.43	2.16	Qatar	104	301.12	2.90	Japan	107	269.14	2.52	Qatar	105	289.37	2.76
Portugal	83	159.83	1.93	Qatar	85	251.68	2.96	Hong Kong	94	342.82	3.65	Portugal	92	183.49	1.99	Kuwait	103	238.24	2.31

Table 6
Money Laundering Attractiveness Index

2007				2006				2005				2004				2003				2002			
Country	Co unt	Sum	Average	Country	Co unt	Sum	Average	Country	Co unt	Sum	Average	Country	Co unt	Sum	Average	Country	Co unt	Sum	Average	Country	Co unt	Sum	Average
Switzerland	136	896.57	6.59	Switzerland	137	907.32	6.62	Switzerland	137	879.99	6.42	Switzerland	137	923.66	6.74	Luxembourg	135	789.99	5.85	Luxembourg	135	2,945.69	21.82
Luxembourg	135	812.28	6.02	Canada	137	373.08	2.72	Luxembourg	136	703.34	5.17	Luxembourg	136	714.13	5.25	Japan	135	409.94	3.04	Japan	134	394.66	2.95
United States	135	467.30	3.46	Luxembourg	136	735.27	5.41	Denmark	136	731.09	5.38	United Kingdom	136	485.54	3.57	Austria	135	487.60	3.61	Switzerland	134	720.56	5.38
Denmark	135	800.38	5.93	Denmark	136	789.09	5.80	Ireland	136	472.37	3.47	Denmark	136	728.53	5.36	Switzerland	135	965.30	7.15	Austria	134	331.76	2.48
United Kingdom	135	527.14	3.90	Ireland	136	531.98	3.91	Netherlands	136	502.00	3.69	Netherlands	136	526.40	3.87	United Kingdom	135	490.89	3.64	Canada	134	313.37	2.34
Ireland	135	547.52	4.06	United Kingdom	136	496.11	3.65	Iceland	136	587.96	4.32	Sweden	136	394.42	2.90	Denmark	135	745.99	5.53	Netherlands	134	381.79	2.85
Sweden	135	422.47	3.13	Netherlands	136	489.42	3.60	Cyprus	135	463.97	3.44	Ireland	136	432.69	3.18	Netherlands	135	544.45	4.03	Ireland	134	293.53	2.19
Netherlands	135	472.47	3.50	Iceland	136	736.65	5.42	United Kingdom	135	467.38	3.46	Germany	136	421.58	3.10	Sweden	135	406.15	3.01	Denmark	134	293.29	2.19
Iceland	135	607.32	4.50	Austria	135	445.12	3.30	Sweden	135	393.02	2.91	Iceland	136	416.42	3.06	Germany	135	462.28	3.42	Cyprus	133	317.98	2.39
Cyprus	134	496.08	3.70	Finland	135	272.25	2.02	Austria	134	443.56	3.31	Cyprus	135	465.86	3.45	Ireland	135	377.57	2.80	United Kingdom	133	438.48	3.30
Austria	134	422.29	3.15	Sweden	134	406.16	3.03	Germany	132	378.77	2.87	Austria	135	456.42	3.38	Iceland	135	341.63	2.53	Qatar	132	459.28	3.48
Finland	134	265.66	1.98	United States	133	445.93	3.35	Canada	130	340.78	2.62	Moldova	134	439.84	3.28	Cyprus	134	489.11	3.65	Germany	128	264.36	2.07
Spain	133	343.04	2.58	Germany	132	372.18	2.82	Mongolia	130	320.01	2.46	United States	134	459.62	3.43	United States	134	465.48	3.47	Belgium	121	341.98	2.83
Germany	127	323.21	2.54	Spain	131	330.34	2.52	United States	127	438.53	3.45	Canada	130	342.59	2.64	Canada	132	356.35	2.70	France	119	228.11	1.92
France	124	264.65	2.13	France	129	280.20	2.17	Spain	125	287.55	2.30	Japan	125	367.99	2.94	France	127	300.01	2.36	Kuwait	117	282.73	2.42
Japan	117	295.78	2.53	Italy	119	254.77	2.14	Finland	122	231.50	1.90	France	118	272.66	2.31	Finland	121	230.48	1.90	Hong Kong	111	382.61	3.45
Kuwait	111	245.86	2.21	Japan	115	318.38	2.77	Japan	117	322.97	2.76	Finland	116	221.79	1.91	Spain	112	248.65	2.22	Finland	109	167.10	1.53
Italy	109	226.00	2.07	Qatar	102	286.68	2.81	France	115	252.46	2.20	Spain	113	250.95	2.22	Kuwait	109	265.08	2.43	Sweden	104	170.92	1.64
Canada	99	226.33	2.29	Kuwait	94	216.84	2.31	Italy	100	222.04	2.22	Kuwait	103	248.57	2.41	Italy	109	259.49	2.38	Italy	103	207.14	2.01
Qatar	95	264.52	2.78	Portugal	87	163.19	1.88	Kuwait	87	208.74	2.40	Italy	101	233.84	2.32	United Arab Emirates	95	247.68	2.61	Israel	101	218.57	2.16



The Index indicates that between 2002 and 2022, the most prosperous and most developed nations have consistently placed among the top locations for laundering money. Countries such as Luxembourg, Switzerland, Hong Kong, Qatar, Austria, Sweden, Denmark, Cyprus, as well as the UK, USA, Canada and Japan provide the type of finance infrastructure that criminals want, stable economies, modern banks, as well as the convenience of moving money without examination.

Luxembourg is long regarded as an important financial center and has been the subject of numerous concerns in FATF assessments (2009, 2023) due to its extensive insurance and finance sectors with weak enforcement, as well as the risk of being a target for foreign criminals. In the same way, Switzerland's banking secrecy is still a major draw for money laundering gangs, despite changes. In the year 2020, Swiss authorities confiscated \$10 million worth of illegal gold tied to international networks.

Large financial centers like Japan, Hong Kong, the US and the UK can also attract illicit funds. The influx of foreign direct investment (FDI) can be used as a cover to facilitate laundering, and financial institutions gain from the flow of low-cost deposits. Even stable economies such as Sweden or Denmark. In addition, Austria attracts such activity because of its low tax rates, the strength of its service industries, and its proximity to trafficking routes.

Qatar is under scrutiny for terror financing and money laundering dangers. Additionally, offshore havens such as Cyprus and Panama have a plethora of appeal due to loosen regulations and fast company formations. The open system of immigration in Canada has brought up similar issues.

This global phenomenon, in which dirty money derived from weak governance systems in the developing world gets deposited in more developed economies, has been well-documented (Levi and Reuter, 2006, and Quirk, 1997). Many countries that are thought to be secure, such as Sweden and Finland, have now been exposed to money laundering risk, which underscores the need for greater international oversight and cooperation to combat money laundering.

Table 7
 List of Countries with Number of Times Appeared from 2002 to 2022

Countries	No. of Times	Score Total	Average	Countries	No. of Times	Score Total	Average	Countries	No. of Times	Score Total	Average	Countries	No. of Times	Score Total	Average
Switzerland	2937	18,883.30	6.43	Korea, South	599	1,572.26	2.62	Bulgaria	61	86.12	1.41	Bhutan	4	6.64	1.66
Luxembourg	2917	26,794.59	9.19	Malta	576	1,062.69	1.84	Libya	61	104.30	1.71	Azerbaijan	3	7.29	2.43
Denmark	2891	13,101.94	4.53	Czech Republic	561	817.57	1.46	Mauritius	55	116.08	2.11	Belarus	3	3.32	1.11
Netherlands	2878	9,246.20	3.21	Bahamas, The	478	1,703.44	3.56	Honduras	45	41.90	0.93	Iran	2	2.43	1.21
Sweden	2846	8,278.46	2.91	Brunei	407	1,177.63	2.89	China	41	50.94	1.24	Peru	2	2.46	1.23
United Kingdom	2843	9,058.92	3.19	Bahrain	387	1,342.03	3.47	Turkey	39	56.55	1.45	Senegal	2	2.27	1.14
Germany	2761	6,655.58	2.41	Slovakia	358	526.78	1.47	Belize	33	64.26	1.95	Ecuador	1	1.12	1.12
Austria	2758	8,127.96	2.95	Hungary	308	432.92	1.41	Jordan	31	75.42	2.43	Haiti	1	1.55	1.55
Finland	2755	5,228.29	1.90	Panama	271	609.06	2.25	Romania	30	48.72	1.62	Indonesia	1	1.25	1.25
Ireland	2742	7,536.51	2.75	Trinidad & Tobago	258	829.85	3.22	Dominican Republic	24	66.66	2.78	Mexico	1	1.33	1.33
Cyprus	2741	9,401.20	3.43	Croatia	249	418.82	1.68	Bosnia & Herzegovina	23	19.08	0.83	Nicaragua	1	1.18	1.18
Iceland	2676	6,639.70	2.48	Malaysia	224	536.69	2.40	Colombia	23	35.13	1.53	Papua New Guinea	1	1.63	1.63
France	2574	5,421.64	2.11	Poland	221	236.97	1.07	Suriname	22	61.78	2.81	Philippines	1	1.32	1.32
Japan	2406	6,369.92	2.65	Grenada	190	659.58	3.47	Argentina	20	36.51	1.83	Russia	1	1.37	1.37
Qatar	2369	7,150.58	3.02	Oman	148	326.76	2.21	Albania	19	28.93	1.52	Solomon Islands	1	1.37	1.83
United States	2262	7,134.88	3.15	Moldova	136	444.15	3.27	Fiji	19	34.49	1.82	Vietnam	1	1.16	1.16
Spain	2201	4,641.92	2.11	Mongolia	135	343.47	2.54	Jamaica	17	24.31	1.43				
Italy	2100	4,404.80	2.10	Costa Rica	133	291.15	2.19	Botswana	16	23.80	1.49				
Hong Kong	2031	8,929.35	4.40	South Africa	128	1,649.37	12.89	Seychelles	16	25.10	1.57				
Canada	1880	4,360.16	2.32	Saudi Arabia	123	226.69	1.84	Armenia	15	23.24	1.55				
Kuwait	1763	4,096.02	2.32	Chile	118	427.00	3.62	El Salvador	15	23.35	1.56				
Singapore	1676	6,312.02	3.77	Uruguay	111	287.28	2.59	Macedonia	15	21.06	1.40				
Belgium	1602	4,053.14	2.53	Thailand	105	161.34	1.54	Georgia	13	24.08	1.85				
Greece	1507	3,113.34	2.07	Lebanon	91	197.20	2.17	Guyana	13	23.87	1.84				
United Arab Emirates	1497	3,890.41	2.60	Lithuania	87	141.06	1.62	Samoa	11	16.77	1.52				
Israel	1386	3,293.40	2.38	Dominica	84	190.28	2.27	Tunisia	10	14.05	1.40				
Portugal	1369	2,523.86	1.84	Venezuela	83	141.73	1.71	Sierra Leone	8	14.62	1.83				
Australia	1299	3,706.75	2.85	Latvia	82	136.58	1.67	Vanuatu	7	9.93	1.42				
Macau	735	3,461.78	4.71	Serbia	82	137.80	1.68	Maldives	5	7.82	1.56				
Slovenia	671	1,356.64	2.02	Brazil	66	145.72	2.21	Togo	5	9.42	1.88				



Table 08 presents countries that are not highly ranked as preferred destinations for money laundering for a number of reasons summarized here. Nations and regions under conflict, political instability, and economic strife are typically shunned by investors in search of a safe haven for their money. The same would hold for criminals as they keep their dirty money out of regions that are politically unstable and thus put their funds at risk. These areas lack the requisite stability and infrastructure for a secure environment in which to preserve wealth, and thus, are generally unattractive markets for money launderers. This issue is more frequent in less-developed countries than in developed countries. Developing countries lost USD 7.8 trillion between 2004-2013, transferred to developed countries with USD \$1.1 trillion in 2013 alone (Global Financial Integrity, 2015). There is also less access to banking and financial services in these areas, which poses challenges for criminals attempting to launder money as they can't easily locate legitimate-seeming outlets for these funds. Small economies are also a source of challenges — fewer transactions and financial activity make illicit money harder to hide. Hand-rich-shoulder-deep-pocket-few-pockets with guano, there aren't as many places to hide ill-gotten booty and therefore a greater chance of detection. This combination of conflict and instability, small economic size and few financial services makes it unattractive to money launderers, who would rather seek out jurisdictions that are stable and developed. This conclusion is also shared by researchers such as Head (2003) and Unger et al. (2006), and Khan et al. (2018).

Table 08
List of Countries with Zero Appearance from 2002 to 2022

Afghanistan	Kazakhstan
Algeria	Kenya
Bangladesh	Kyrgyzstan
Benin	Laos
Bolivia	Lesotho
Burkina Faso	Liberia
Burundi	Madagascar
Cambodia	Malawi
Cameroon	Mali
Cape Verde	Mauritania
Central African Rep.	Nepal
Chad	Niger
Comoros	Nigeria



Congo, Dem. Rep.	Pakistan
Congo, Repub. of the	Paraguay
Cote d'Ivoire	Sri Lanka
East Timor	Sudan
Egypt	Swaziland
Gambia, The	Tajikistan
Ghana	Tanzania
Guinea	Uganda
Guinea-Bissau	Ukraine
India	Yemen
Iraq	Zambia

Below is a trend analysis, for the purpose of this analysis, we have listed down the top 10 countries from each year from 2002 to 2022, and the list comprises of nineteen common countries, which are sorted by average of count. In order to make the graph simpler and presentable, the list has been divided into two sections⁵. The first section consists of ten countries where whereas the latter part comprises of nine countries. The first section with more preferred destinations shows a rather consistent trend, as shown in Graph 1. However, Graph 2 shows that countries in the second section have a mixed and erratic trend. This breakdown identifies that for almost two decades, preferred destinations for money laundering have remained the same.

⁵ List attached in Appendix



Figure 1
Trend Analysis of Top 10 Countries

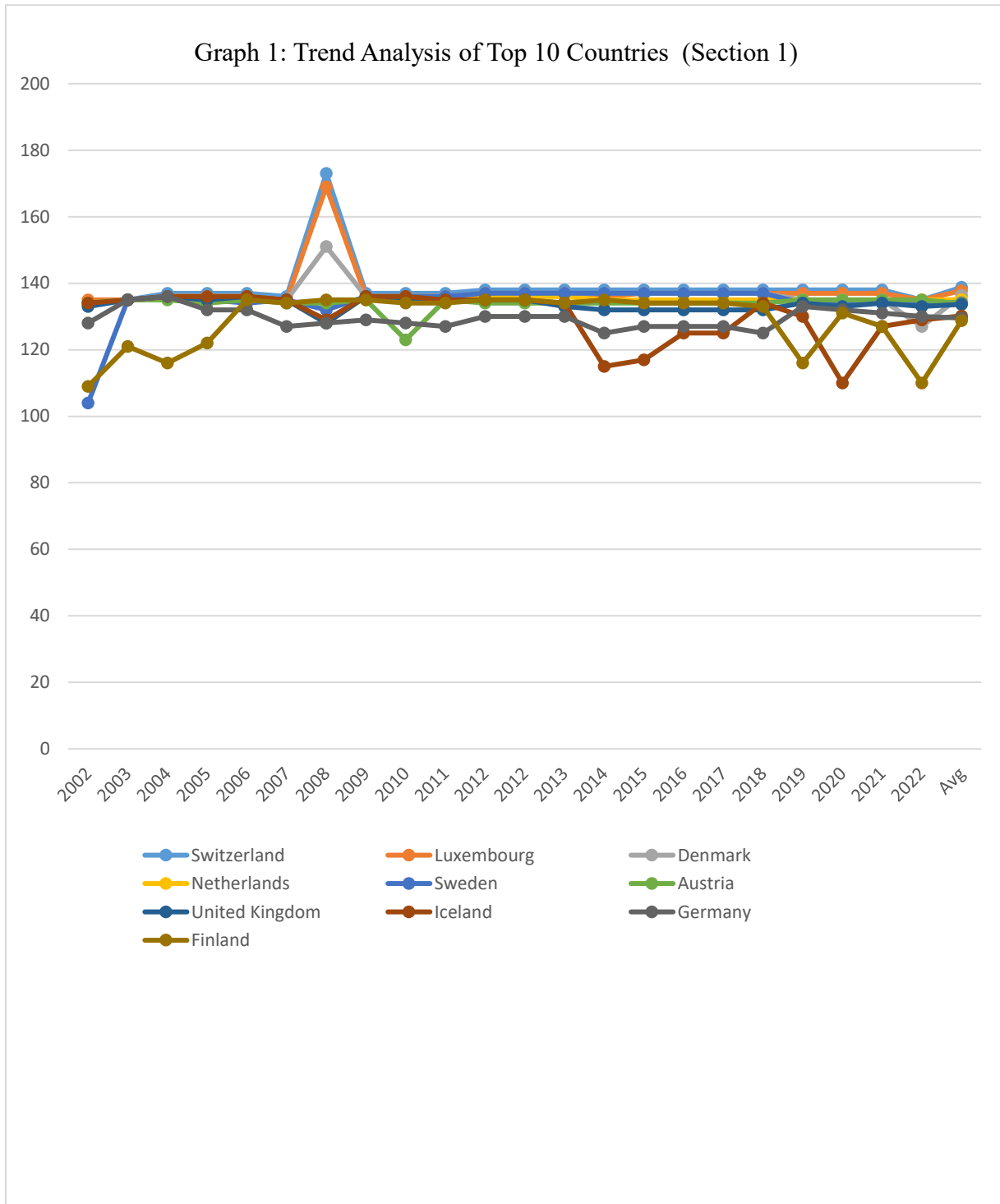
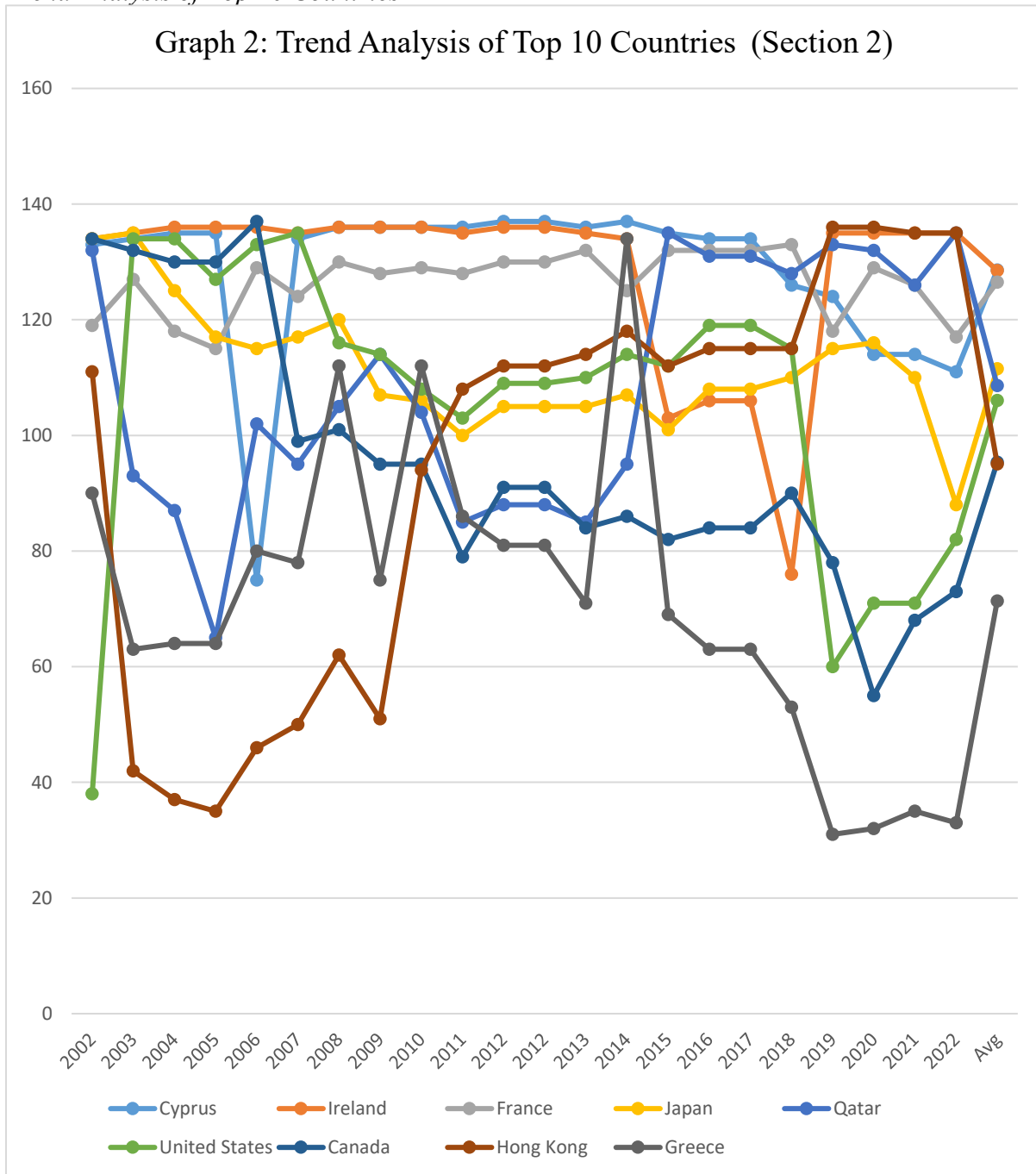




Figure 2
Trend Analysis of Top 10 Countries





Results based on the Countries classification

The World Bank's income-based classification of 2022, that is, low, lower-middle, upper-middle, and high income, is utilized in this paper to examine money laundering dynamics by development level on the basis of GNI per capita. GNI is a valid indicator for the economic potential and growth opportunities of a country, consistent with earlier studies that have highlighted the importance of income as a driver of human development (Sen, 1999; Nielsen, 2011). When we segment countries in this way, we identify 31 distinct jurisdictions (among the 25 top laundering destinations for each group of countries). The uppermost of the list is topped by high-income countries such as Luxembourg, Switzerland and Denmark. High-risk financial services and cross-border transactions are provided in Luxembourg with continued non-financial risks despite the measures taken in the area of AML. Switzerland's banking history continues to be buoyant, despite recent scandals such as those involving Credit Suisse that highlight enduring risks. Financially, Denmark is stable, but it is also high risk for money laundering given that its economy is relatively open and it is strategically located. This classification based on GNI is not only

Table 9
List of 31 Common Favorite Destinations for Money Laundering from each Classification Group

High Income Countries (> 13,845)				Upper Middle Countries (4,466 - 13,845)				Lower Middle Countries (1,136 - 4,465)				Low Income Countries (<= 1,135)			
Country	No. of Times	Score Total	Average	Country	No. of Times	Score Total	Average	Country	No. of Times	Score Total	Average	Country	No. of Times	Score Total	Average
Luxembourg	1014	10902.32	10.75	Luxembourg	786	6293.37	8.01	Luxembourg	822	7004.47	8.52	Luxembourg	295	2594.43	8.79
Switzerland	1030	7300.96	7.09	Switzerland	784	4669.19	5.96	Switzerland	821	4917.99	5.99	Switzerland	302	1995.16	6.61
Denmark	1004	5440.35	5.42	Denmark	773	3105.99	4.02	Denmark	819	3255.99	3.98	Denmark	295	1299.61	4.41
Netherlands	1002	4049.82	4.04	United States	601	3006.58	5.00	Cyprus	781	3201.06	4.10	Cyprus	291	1029.64	3.54
United Kingdom	979	3670.49	3.75	Cyprus	702	2591.21	3.69	Hong Kong	674	3128.29	4.64	South Africa	50	974.11	19.48
Austria	967	3641.27	3.77	Singapore	457	2288.28	5.01	United Kingdom	811	2433.70	3.00	Netherlands	293	935.07	3.19
Sweden	990	3324.82	3.36	Hong Kong	559	2065.33	3.69	Japan	749	2356.57	3.15	United Kingdom	292	890.21	3.05
Hong Kong	592	3232.45	5.46	United Kingdom	761	2064.52	2.71	Singapore	624	2330.84	3.74	Ireland	284	886.05	3.12
Germany	978	2887.08	2.95	Netherlands	770	2048.81	2.66	Netherlands	813	2212.50	2.72	Sweden	290	796.71	2.75
Ireland	961	2882.38	3.00	Sweden	763	2031.06	2.66	Qatar	767	2193.09	2.86	Austria	279	741.13	2.66
Qatar	748	2875.30	3.84	Austria	738	1934.43	2.62	Sweden	803	2125.87	2.65	Iceland	273	736.87	2.70
Cyprus	967	2579.29	2.67	Iceland	718	1873.29	2.61	Ireland	755	1903.12	2.52	Qatar	285	733.24	2.57
Belgium	681	2317.94	3.40	Ireland	742	1864.96	2.51	Iceland	748	1828.37	2.44	Spain	261	697.43	2.67
Iceland	937	2201.17	2.35	Japan	678	1803.18	2.66	Austria	774	1811.12	2.34	France	279	668.64	2.40
France	911	2076.43	2.28	Canada	541	1609.49	2.98	Australia	477	1718.97	3.60	United States	265	601.97	2.27
Finland	986	2013.03	2.04	Germany	744	1583.83	2.13	United States	663	1682.15	2.54	Italy	278	587.90	2.11
United States	733	1844.18	2.52	Qatar	569	1348.95	2.37	Germany	773	1628.78	2.11	Germany	266	555.90	2.09
Japan	703	1706.49	2.43	Finland	726	1325.95	1.83	Kuwait	626	1457.35	2.33	Japan	276	503.67	1.82
Italy	814	1677.51	2.06	France	698	1295.16	1.86	Israel	435	1449.94	3.33	Hong Kong	206	503.27	2.44
Macau	200	1669.82	8.35	Australia	443	1123.98	2.54	Finland	765	1389.77	1.82	Finland	278	499.55	1.80
Spain	814	1634.49	2.01	Spain	573	1111.03	1.94	France	686	1381.41	2.01	Kuwait	238	478.12	2.01
UAE	401	1443.46	3.60	Bahamas, The	294	1100.23	3.74	UAE	574	1322.47	2.30	Portugal	209	477.19	2.28
Canada	608	1279.98	2.11	Italy	491	1041.59	2.12	Spain	553	1198.96	2.17	Singapore	203	439.92	2.17
Singapore	392	1252.97	3.20	Kuwait	380	944.33	2.49	Canada	537	1106.69	2.06	UAE	186	393.57	2.12



Kuwait	519	1216.21	2.34	Greece	316	935.49	2.96	Italy	517	1097.80	2.12	Israel	193	384.98	1.99
Greece	639	1054.70	1.65	Belgium	416	763.14	1.83	Macau	296	1082.17	3.66	Greece	199	380.88	1.91
Portugal	547	858.83	1.57	UAE	336	730.91	2.18	Greece	353	742.27	2.10	Canada	194	363.99	1.88
Israel	463	799.55	1.73	Israel	295	658.92	2.23	Belgium	358	660.21	1.84	Belgium	147	311.85	2.12
Australia	272	652.90	2.40	Macau	194	613.67	3.16	Portugal	292	649.50	2.22	Australia	107	210.90	1.97
Bahamas, The	119	385.74	3.24	Portugal	321	538.34	1.68	South Africa	54	418.48	7.75	Macau	45	96.11	2.14
South Africa	0	0.00	0.00	South Africa	24	256.78	10.70	Bahamas, The	65	217.47	3.35	Bahamas, The	0	0.00	0.00



The rest of the list consists of countries which deemed as global and regional financial hubs providing services mainly to foreign clientele through facilitation and ease in establishing offshore companies, zero or low-cost movement of funds across borders. Countries establish the list constitutes jurisdictions that are the major recipients of foreign direct investments in the world. It is suggested that 20% of foreign direct investment motives are illegal outflow of funds, and FDI is used as a tool for money laundering (Perez & Drabek, 2012). The table below shows the list of the top 25 countries with the highest net inflow of FDI out of 25 countries; 17 countries are common with the 31 countries mentioned table.

Table 10 List of Countries that received the Highest FDI		
United States	Canada	Malta
China	British Virgin Islands	Japan
Netherlands	Spain	Cyprus
United Kingdom	Australia	Luxembourg
Hong Kong	Belgium	Sweden
Germany	India	
Singapore	Mexico	
Brazil	Russian Federation	
Ireland	Italy	
France	Hungary	

World Bank Data 2002 to 2021 (Average)

Furthermore, countries with a strong financial sector and a better standard of living also attract criminals and corrupt public office holders for laundering and placement of funds for future use. At present many wrong doers do not particularly follow the money laundering process to bring back the illegal money to their source countries to avoid the risk of confiscation rather they keep such funds abroad by making up such funds look legitimate through complex set of transactions and off shore companies, in many jurisdictions there are various service providers including financial accountants and legal consultants which remain available to assist criminals. Walker & Unger (2009) state that countries with a robust services industry, including financial services and other services, are some of the preferred destinations. The study suggests the triangulation of estimations from Walker's formula with services export statistics. Below mentioned list in table, validates that the countries with strong services sectors, including financial services, tourism, and gambling, are more attractive for money laundering. List is comprising of top thirty-one countries from above estimations and shows strong services sector



in composition to their respective GDP in all cases except those where the country is major oil producing nation, from finding it can easily be said that strong services sector and especially financial services sector show capacity to support money laundering which is an important part of attractiveness.

According to Know Your Country, 780,000 tourists visit Iceland, services sector, including tourism, is also making the country attractive to money launderers. The casino industry, which is an integral part of Iceland’s economy, is also used by money launderers to make their black money white. Due to the cash-based nature of the business, it is not easy to trace back the source of proceeds. According Crime and Misconduct Commission (2005) chips are bought through cash and whether in gambling places or not, and chips are cash back possibly through cheques favoring a third party, even chips are also used to purchase valuable goods. Cyprus is an eastern Mediterranean island which has dominated by the services industry, with a composition 87.1% of GDP according to the CIA World Factbook. Mainly, banking business relies on business introducers called Administrative Service Provider (ASP), and the normally beneficial owner does not meet with bank authorities directly and in person, which construct major chances of financial sector inclusion in money laundering activities and also a violation of international regulatory requirements.

Table 11
Services Industry composition of GDP in Percentage from top 31
Countries⁶

Hong Kong	91.00%	Greece	67.60%
Macau	90.10%	Portugal	67.00%
Bahamas, the	80.80%	Denmark	66.10%
Luxembourg	80.60%	Sweden	65.30%
United States	76.70%	Italy	64.90%
Cyprus	74.00%	Iceland	64.60%
United Kingdom	72.80%	Australia	64.20%
Singapore	72.40%	South Africa	63.00%
Israel	72.40%	Germany	62.60%
Switzerland	71.90%	Austria	62.30%
Japan	71.40%	Finland	60.70%
Belgium	70.70%	Ireland	56.60%
Canada	69.60%	UAE	47.70%
Netherlands	69.30%	Kuwait	43.60%

⁶ Data collected from CIA Factbook



France	69.20%	Qatar	38.50%
Spain	68.50%		

In short, the list provided in Tables 03 to 06 consists of countries that are proven to be attractive due to several reasons, including a favorable environment for moving funds into and out of the country, or jurisdictions with a strong and vast financial sector, major FDI flows, and lax tax regulations. The developed part of Europe is also concentrated in the list, which shows that wrongdoers prefer such developed parts of the world. Transparency International's report suggests that US\$ 3.7 billion in corruption links were found hidden in wealthy nations in a complex web of offshore accounts, real estate, and shell companies.

Conclusion and Discussion

Money laundering is the lifeblood of organized crime; it is how criminals wash dirty money and reap the profits while looking clean. It is also closely associated with terrorism financing and proliferation financing and is considered to be a significant global issue. Underdeveloped, poor nations suffer from weak governance, insufficient resources, and frail institutions, thereby increasing their exposure to such crimes. Stealing and looting valuable national resources through corruption, fraud and extortion, and taking them elsewhere is the goal of such wrongdoers. This wealth is typically laundered in the context of sophisticated financial systems and is ultimately placed in developed countries, where it becomes legal, protected from seizure and available for use in the future.

The issue endures, despite efforts over the years, ranging from research to laws and anti-money laundering (AML) regimes. Criminals are also advancing and finding new ways to avoid global surveillance. Ironically, the implications are aimed at weaker nations that are the source of money launderers. But this research confirms a different reality: instead of benefactors, these nations are losers. The money flees these countries, worsening challenges such as poverty, debt, slow growth, and underfinanced education and health systems.

While developed nations preach AML, they often act as the end recipients of dirty money. Their highly developed financial sectors, excellent quality of life, and legal protections are a draw for criminals who want a safe place and a veneer of legitimacy for their money. Poor international cooperation, absence of bilateral arrangements and gaps, such as anonymous shell companies, continue to give impetus to this phenomenon. This unequal enforcement of



AML rules has created an imbalance on a global scale; poor countries are subjected to great scrutiny, while rich ones too frequently aid the laundering they pretend to abhor.

This article introduces a new yardstick for gauging the most appealing spots for laundered money, a powerful tool for policymakers to gain insight into where the money is going. If aggregated efforts to fight money laundering are to be meaningful, the world needs an independent, empowered global institution with investigative and regulatory authority. Such an entity should also operate internationally, question suspicious wealth even when it looks legal, and ensure that regulatory standards are enforced wherever you are.

At its core, money laundering is not simply a financial crime, it is a global plague. It sustains terrorist networks and drug- and human-smuggling operations, and it destroys faith in institutions everywhere. Without a comprehensive, transparent and enforceable global response, attempts to combat it are bound to fail. The study also urges a united, comprehensive global coalition to close the loopholes, repatriate stolen assets, and bring the financial criminals to justice wherever in the world they may be holding.

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