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► To cite this version:

Nikola Kutin, Patrice Guillotreau, Thomas Vallée. Assessing competition on Maritime Routes in the Liner Shipping Industry through multivariate analysis. 2018. <halshs-01828643>

HAL Id: halshs-01828643

<https://halshs.archives-ouvertes.fr/halshs-01828643>

Submitted on 3 Jul 2018

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Assessing competition on Maritime Routes in the Liner Shipping Industry through multivariate analysis

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2018/09

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Assessing competition on Maritime Routes in the Liner Shipping Industry through multivariate analysis

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Abstract

The current paper investigates the level of competition on maritime routes in the liner shipping industry by applying multivariate and cluster analyses on maritime indicators. We use a dataset which includes maritime routes between 153 ports for the year 2014, described by several characteristics regarding the number of operators, the number of ships and trips, the size of ships, the sea distance, the bilateral countries' connectivity. Some clusters of maritime routes are identified along two key components, a first one related to the number of competing firms, and a second one where the average size of firms is positively correlated with distance. The first one indicates somehow the degree of competition while the second one is related to the efficiency of carriers. Another way of looking at competition is to consider the region-based trade and to see whether indicators respond differently from region to region.

Keywords: multivariate analysis, clusters, competition, shipping.

This work was supported by the European Union's Erasmus+ Programme, Key Action 2, Capacity building in the field of higher education under DOCKSIDE Project (www.dockside-kh.eu), Grant number: 573790-EPP-1-2016-1-FR-EPPKA2-CBHE-SP.

1. Introduction

Since the invention of the shipping container by Malcolm McLean in 1956, the containerized trade has experienced a remarkable growth. It allowed port operators and shipping carriers to reduce their loading and unloading costs and to considerably improve their time efficiency. According to Clarkson Research's data, since 1990, the container trade has increased by more than 600%. In 2016, it accounted for only 16.7% of the total seaborne trade, while its value was more than 60%¹. In the last decades, we have observed a specialization of the ports by investing in the construction and enlargement of container terminals and connections with the hinterland network. Major investments, such as the \$8.2 billion expansion program of the Suez Canal completed in July 2015² and the enlargement of the Panama Canal for \$5.25 billion achieved in June 2016, made faster and cheaper the operation of large container vessels.

From 1999 to 2016, the average size of containerships and the volume of containerized cargo per mile has increased by 127% and 208%³, respectively. This positive trend is largely driven by the rapid pace of globalization which was amplified by the inclusion of China in the WTO membership in December 2001. As a consequence of globalization, the prevailing network structure in the maritime trade has turned into a "hub and spoke" structure. Some ports, such as Hong Kong and Singapore, located on central strategic geographical sites (hubs) have direct connections with regional ports (feeders or spokes) and other hub ports.

Along this hub and spoke network organization, the liner shipping industry has been rapidly developed by exploiting large increasing returns to scale, through which the most transited

¹ <http://www.worldshipping.org/about-the-industry/global-trade>

² <https://www.economist.com/news/middle-east-and-africa/21660555-it-necessary-bigger-better-suez-canal>

³ Data from Clarkson Research

routes are covered by the largest ships, having a high loading factor and organized within alliances (Yang *et al.*, 2011). Such a growing average size of ships on a few major routes linking port hubs has reduced the number of liner companies after several waves of mergers & acquisitions (Clark *et al.*, 2004). For market leaders, such an external growth strategy represents a way of defending their positions and increasing entry barriers, like the recent Maersk Line's acquisition of Hamburg Süd and the takeover of Neptune Orient Lines by CMA-CGM in 2017. Furthermore, the excessive capacity in the industry has led to further market consolidation through consortia and alliances (Agarwal and Ergun, 2010, Panayides and Wiedmer, 2011). In 2018, three Mega Shipping Alliances include the ten biggest container lines in the world, collectively accounting for 79% of the global container market⁴. Secondary maritime routes with lower trade volumes are covered by smaller companies operating smaller vessels (Clark *et al.*, 2004). Despite the process of concentration and alliances, container freight rates remain at very low levels, and competition on various trade routes has even intensified (UNCTAD, 2017). One of the reasons lies in the imbalance between supply and demand which has raised shippers' bargaining power benefiting from the upsizing race from shipping companies.

The objective of this research is to analyze the competition on the maritime routes by addressing the following research questions. What is the degree of competition on the different maritime routes? How do the sea distance, bilateral countries connectivity as well as regional trade direction influence the level of competition between liner shipping carriers? How can the different maritime routes be classified in terms of competition, and to what extent the geographical location of ports matters in this typology of routes? To answer these questions, multivariate and cluster analyses have been used on a sample of 153 container ports in 50 countries.

The paper is structured as follows. Firstly, a review of literature on competition between container carriers is provided. Secondly, the methodology related to multivariate analysis along with information about the dataset and the used variables are depicted. Thirdly, the results as

⁴ <https://www.xeneta.com/blog/shipping-alliances>

well as a discussion of the outcomes of this research are shown. Finally, the conclusion of the main findings and some proposals for further research are developed.

2. Literature Review

Competition can have different theoretical meanings: more freedom for firms (free entry or exit in the market), an increasing number of rivals, a move away from collusion towards more independent behaviors, or the reward to obtain, or the penalty, etc. (Vickers, 1995). When applied to transportation, competition is narrowly combined with cooperation – forming by contraction the concept of “coopetition” (Dagnino and Rocco, 2009) to cope with the network constraints and the interdependence of maritime routes as sub-markets. By many aspects, the market conditions of shipping services look like those observed in the airline industry, where the theoretical conditions for perfect contestability are far from being satisfied (Hurdle et al. 1989, Notteboom, 2002): no sunk cost of entry for entrants, same post-entry costs for incumbent and entrants, etc. However, competition is stiff enough to keep freight rates at low levels in spite of the ongoing M&A and concentration process.

Due to this competitive regime, ocean container carriers have continuously sought to maximize their market share and/or minimize their running costs (Song, 2002). Clark *et al.* (2004) demonstrated that directional imbalance in trade between countries, which implies that many carriers are forced to haul empty containers back, have a positive effect on the cost of shipments, leading shipping companies to coalesce within conferences or capacity-sharing agreements. The study also showed that maritime conferences have been exerting some mild monopoly power, adding around 5% to transport costs. However, other studies have found that conference outsiders have increased their market share on a few major routes and that a great proportion of service contracts did not even use the official tariffs of shipping conferences, even though companies were members (Cariou (2008), Cariou and Wolff (2006)). This would explain the repeal of the shipping conference exemption by the European Union in the late 2000s (Global Insight, 2005).

The cooperation between the top 20 ocean shipping companies was analyzed by Panayides and Wiedmer (2011). Their results show that focal members of an alliance maintain preferred relationships for service agreements that are adjusted on a continuous basis. The “global alliances” can hardly be seen as closed “entities”. Ha and Seo (2017) used a panel data model to determine to what extent freight

rates, bunker fuel prices, scale economies and chartered vessel ratios had affected the profits of major shipping carriers. It was found that the route specialization does not necessary influence the companies' profit. Hirata (2017) applied a similar approach to estimate the effect of the Hirschmann-Herfindahl Index (HHI) on container freight rates for a sample of six major container liner shipping routes. Results suggest that higher concentration level does not lead to higher prices, and that the container liner shipping market is rather contestable. Therefore, alliances do not hamper competition, but would rather represent solutions to lower unit operating costs. Concentration might nonetheless have more significant impact on freight rates on secondary markets such as the south-American routes, even though increasing rates can also be explained by demand factors (Sanchez and Wilmsmeier, 2011). This is confirmed by other studies reporting that peripheral markets are more subject to the influence of concentration than main ones: only large carriers can enter and challenge the market positions of incumbents which manage to create entry barriers for smaller outsiders. Sys (2009) analyzed the degree of concentration by estimating the following coefficients: the HHI, the Lorenz curve and the Gini coefficient as well as the Hymer–Pashigan index of market share instability. This study shows first how the global shipping market has concentrated tremendously within a decade (2000-2009), the cumulated share of the top-10 companies passing from 38 to 60%, and the Gini coefficient, yet very high, gaining 10 points, from 0.66 to 0.77. Secondly, two groups of maritime routes were identified. The first one was characterized as a loose oligopolistic market which includes large trade lanes (e.g. transatlantic and transpacific trade for a total trade of 41,000,000 TEU), while the second one is a tight oligopoly which includes new/growing/relatively small container trade routes (e.g. Mediterranean—North America, 1,000,000 TEU volume).

A container industry-specific real options investment model of oligopolistic competition taking into account endogenous price formation in the second-hand vessel market, fuel-efficient investment and endogenous lead times was developed by Rau and Spinler (2016). The outcomes of the study demonstrate that an increasing number of players (moving from monopoly to oligopoly) results in higher optimal capacities, lower individual firm values as well as earlier investment. However, an increase in competitive intensity was deemed to reduce optimal capacity and firm value. Wang et al. (2014) concluded that the expansion of the fleet capacity is less costly than updating the frequency of the required services. An additional incentive for cooperation might be the fact that the “grand coalition’s” profit is always higher than the sum of “subcoalition” ones (Liu *et al.*, 2016).

In the literature, many variables were used to analyze the nature and the degree of competition in the container trade. The geographical locations (Anderson et al. (2008); Fraser et al. (2016); Yap et al. (2006); De Oliveira and Cariou (2015); Clark et al. (2004)), the connectivity of the country and the physical infrastructure (Yeo et al. (2008); Fraser et al. (2016)) should be considered. Directional imbalance of the trade was taken into account in the studies of Clark et al. (2004) and Asgari et al. (2013). The increasing size of the fleet and incentives for economies of scale were analyzed in Notteboom and Yap (2012), Ha and Seo (2017) and Rau and Spinler (2016). The oligopolistic nature of the liner shipping industry was shown by Sys (2009). However, we have not found any study classifying different maritime routes on the basis of the variables mentioned above. This paper fills in the gap in the literature by providing a comprehensive analysis of the nature and degree of competition on different maritime routes. Qualitative variables related to the maritime routes such as trade direction, country, region and continent of destination and origin were included in the analysis. We also took into account the bilateral country connectivity as a proxy of the port' infrastructure as well as the number of trips, ships and operators between each pair of ports. We identify similarities between different maritime routes based on this set of variables. Thereafter, by using a Principal Component Analysis and a Cluster Analysis we highlight a typology of trade routes by level of concentration and regional characteristics.

3. Methodology

3.1. Data

In this paper we have used data from multiple sources. A dataset on the ports' connectivity was obtained from Lloyd's List Intelligence⁵. It contains port to port connectivity in 2014 for 153 ports from 50 countries.

Table 1 Distribution of ports and countries in the sample according to the continents

	Countries	Ports
Americas	11	35
Europe	12	20
Asia	18	79
Africa	9	19
Total	50	153

Source: Lloyd's List Intelligence

⁵ <https://www.lloydslistintelligence.com/>

As shown in Source: *Lloyd's List Intelligence*

, most of the ports are located in Asia, followed by Central, Latin and North America, Europe, and Africa. The dataset contains information about 6,410 maritime routes. For each route, we have the following descriptive variables:

- port of departure and port of arrival,
- country, region and continent of departure and arrival,

In addition to these categorical variables, we have the following continuous and discrete variables for each maritime route:

- Average size of container vessels measured in twenty-foot equivalent units (TEU) operating between a port of departure and a port of arrival;
- Number of ships between a port of departure and a port of arrival;
- Number of trips between a port of departure and a port of arrival;
- Number of operators (liner shipping container carriers) operating between a port of departure and a port of arrival;
- Number of ships per operator on a maritime route.

We have also included in our analysis the bilateral sea-distance between the main container ports in each country. It was computed by Bertoli et al. (2016). The database is developed at the country level and, if the two ports belong to the same country, the distance is set at zero. This variable shows the relative length of each maritime route.

Another variable at a country level which we use for the analysis is the Liner Shipping Bilateral Connectivity Index (LSBCI) which was developed by Hoffmann et al. (2014). LSBCI is an extension of UNCTAD's country level Liner Shipping Connectivity Index (LSCI). The LSBCI includes the following components: 1) the number of transshipments required to get from country A to country B; 2) the number of direct connections common to both country A and country B; 3) the geometric mean of the number of direct connections of country A and of country B; 4) the level of competition on services that connect country A to country B; 5), the

size of the largest ships on the weakest route connecting country A to country B (Fugazza and Hoffmann, 2016).

Finally, the descriptive variable Trade Direction is a descriptive variable and allows us to analyze the imbalance of container trade between developed and less developed countries. We use the analytical classification made by the World Bank of the world's economies based on estimates of gross national income (GNI) per capita for 2014⁶. Countries classified as High Income states are considered as Developed (D), while the remaining countries are considered “Less developed” (DL). We obtain therefore four types of trade directions, D-D, LD-LD, D-LD and LD-D.

Table 2 Variables used in this study

Variable Name	Description	Source
Avg_TEU	Average size of container vessels operating on a maritime route measured in TEU	Lloyd's List Intelligence
Ships	Number of ships between a port of departure and a port of arrival;	Lloyd's List Intelligence
Trips	Number of trips between a port of departure and a port of arrival	Lloyd's List Intelligence
Carriers	Number of liner shipping container carriers operating between a port of departure and a port of arrival	Lloyd's List Intelligence
Ships_per_carrier	Number of ships per operator on a maritime route	Computed by authors
Distance	Bilateral sea-distance between the main container ports in each country	Bertoli et al. (2016)
LSBCI	Liner Shipping Bilateral Connectivity Index	Hoffmann et al. (2014)
Trade_direction	Direction of trade according to the national income of the country of departure and the country of arrival.	World Bank

Error! Reference source not found. provides a summary of the variables that we have used for the PCA and Cluster Analysis. It should be highlighted that we decided to restrict the analysis to the long distance routes (more than 5000 km) due to the heterogeneous nature of the observations, we have decided to conduct the Multivariate Analysis on the routes with regular services. Therefore, the routes with less than ten carriers, twenty ships were excluded.

⁶ <https://blogs.worldbank.org/opendata/new-country-classifications-2016>

3.2. Principle Component and Cluster Analyses

Principle Component Analysis (PCA) was introduced by Pearson (1901) and later developed by Hotelling (1933). It is one of the oldest multivariate techniques. PCA allows us to reduce the dimensionality of a data set in which there are a large number of interrelated variables, while retaining as much as possible of the variation present in the data set (Jolliffe, 1986). This methodology is applied to a data table where rows are individuals and columns are variables.

The maritime routes between each pair of ports (port of departure and port of arrival) are considered as active individuals, i.e contributing to the estimated inertia of the cloud of individuals. The following continuous and discrete variables play the role of active elements in the analysis: Avg_TEU, Ships, Carriers, Trips, Distance, Carriers_ships and LSBCI, meaning that the inertia (or variance) of the cloud of variables is calculated on the mere basis of these active variables. Supplementary elements are also included but do not contribute directly to the factor analysis, being simply projected on the factorial maps built up by active elements: Route_Country, Route_Continents, Route_Regions and Trade_Direction play the role of illustrative discrete variables. The supplementary elements make it possible to illustrate the principal components.

We have also divided the maritime routes into different clusters by applying the Hierarchical Agglomerative Clustering Analysis based on prior Principle Components. The distance between the clusters has been computed by the Ward's Method (Ward Jr, 1963).

4. Results

In this part are described the results. The first section provides descriptive statistics related to the whole dataset of 6,410 maritime routes. In the second and third sections are depicted the results of the PCA and cluster analyses for a subset of 800 maritime routes. Finally, the typology

of the maritime routes as well as the absolute concentration in terms of number of ships per carrier are presented.

4.1. Descriptive statistics

By looking first at the whole sample, we analyze the degree of competition on 6,410 maritime routes connecting 153 ports from 50 countries. **Error! Reference source not found.** provide descriptive statistics of the variables used in this study:

Table 3 Descriptive statistics for the whole dataset (6,410 routes)

Variable name	Count	Mean	Std. Deviation	Minimum	Maximum
TEU	6,410	4,299.77	3,015.23	80.00	18,270.00
Ships	6,410	25.20	54.07	1.00	1,237.00
Carriers	6,410	11.14	17.90	1.00	264.00
Trips	6,410	99.70	277.10	1.00	5,510.00
Distance	6,410	6,771.12	5,906.48	0.00	22,374.60
Ships_per_carrier	6,410	1.86	1.21	1.00	19.00
LSBCI	5,630	0.55	0.15	0.19	0.86

As we can see in **Error! Reference source not found.**, the dataset contains heterogeneous maritime routes. There are routes with only one ship operating between the port of origin and the port of arrival. The average size of ships, which is a proxy for economies of scale, also varies significantly: the mean is 4,299.77 and the maximum value is 18,270. On average, there are 25 ships and 99.7 trips per route. The average number of ships per carrier is 1.86, but some carriers can deploy up to 19 ships on a single route. Unfortunately, the distribution of market shares per route between operators, which would allow calculation for classical concentration indicator, was not available in this dataset. The high mean value of LSBCI (0.55) indicates that countries in the sample are relatively well connected between each other.

Table 4 Pearson's correlation matrix of the variables related to 6,410 routes

	TEU	Ships	Carriers	Trips	Distance	ships_per_carrier	LSBCI
TEU	1.000						
Ships	0.128	1.000					
Carriers	0.018	0.921	1.000				
Trips	-0.044	0.604	0.711	1.000			
Distance	0.592	-0.081	-0.165	-0.156	1.000		

Ships_per_carrier	0.407	0.328	0.205	0.165	0.191	1.000	
LSBCI	0.283	0.278	0.242	0.213	0.042	0.227	1.000

There is an expected strong correlation between the number of carriers and that of ships. Also not surprising is the high positive correlation between the latter and the number of trips. However, the positive correlation coefficient between Distance and TEU indicates that the average size of operating ships increases with the distance between the port of origin and destination. Economies of scale would therefore increase with distance and may reduce the number of competing carriers. The number of ships per carrier, which is a proxy for the size of shipping companies, is also higher on the routes where these big vessels operate. Distance between ports would therefore be a key variable of lower competition through higher economies of scale acting as entry barriers.

Regarding the regional container trade, there are some fundamental differences across regions. In the dataset, the maritime routes are well distributed between “Developed” and “Less developed” countries. There are 1,766 routes between Developed countries (D-D), 1,414 routes connecting “Developed” and “Less developed” states (D-LD), 1,447 routes between Less developed and Developed nations (LD-D) and 1,783 routes link exclusively less developed states (LD-LD).

Table 5 Mean values according to the direction of the trade. D refers to “Developed” countries and LD refers to “Less Developed” ones. Developed states are those considered as "high income" countries by the World Bank

	Mean (D-D)	Mean (D-LD)	Mean (LD-D)	Mean (LD-LD)
TEU	5,073.47	4,683.84	4,725.07	2,883.67
Ships	30.52	24.81	24.49	20.82
Carriers	12.86	10.97	10.86	9.81
Trips	145.59	91.71	93.17	65.87
Distance	7,298.05	7,921.28	7,575.73	4,684.00
Ships_per_carrier	2.21	1.80	1.84	1.51
LSBCI	0.61	0.55	0.55	0.48

Error! Reference source not found. reveals the relative imbalance according to the level of income of the countries in the sample. The mean values of all variables are the highest on the maritime routes connecting “Developed” states (DD). It indicates that on these routes, the relative connectivity of the ports is also high and the traffic is the most intensive. The average

size of operators is also greater (2.21 ships per operator), which indicates more concentrated markets.

On the other hand, the containerized trade between "Less developed" (Low income, Lower middle income, and Upper middle income countries) is less concentrated and the average values of all variables are the lowest ones. It is interesting to highlight that the maritime routes NS and SN are not perfectly symmetrical. The average size of ships on the "D-LD" routes are on average smaller and their number is slightly higher than those operating on the "LD-D" routes. This indicates a regional misbalance of the trade going from "Developed" to "Less developed" countries and vice versa. The services on the "LD-D" routes are more frequent and the vessels are smaller. In addition, there are more carriers on the "D-LD" routes. It could be explained by the fact that, often the vessels are not fully loaded on the ports in developed countries due to the low demand in less developed states. As a consequence, some carriers cooperate by forming alliances in order to optimize their costs.

When we look at the trade between continents we observe that the container trade patterns differ considerably.

Route	Mean TEU	Mean Ships	Mean Carriers	Mean Trips	Mean Distance	Mean Ships per carrier	Mean LSBCI
Eur-Asia	8,757.19	25.01	7.25	59.34	14,623.46	2.80	0.62
Asi-Eur	8,488.66	27.99	7.96	73.03	13,708.51	2.76	0.62
Ame-Asia	5,698.62	21.24	9.06	56.06	13,452.31	2.04	0.58
Asia-Ame	5,697.69	24.30	10.28	67.28	12,569.75	2.00	0.59
Afr-Asia	5,474.79	28.33	10.58	65.51	11,881.27	1.68	0.42
Asia-Afr	5,398.96	26.42	11.26	64.41	10,587.49	1.82	0.43
Afr-Ame	4,800.08	9.58	5.14	24.91	10,586.55	1.48	0.47
Eur-Ame	4,469.81	16.49	7.77	59.49	8,805.08	1.99	0.57
Ame-Eur	4,419.10	15.80	7.31	56.27	8,708.18	2.06	0.58
Eur-Eur	4,241.52	54.07	23.23	328.18	2,306.26	2.04	0.66
Ame-Afr	4,206.42	11.60	5.65	30.13	10,785.53	1.49	0.44
Ame-Ame	3,912.69	23.15	11.14	87.91	2,794.57	1.78	0.50
Eur-Afr	3,343.37	23.76	11.20	98.68	3,423.56	1.75	0.46
Afr-Eur	2,969.62	18.10	10.46	87.06	3,495.96	1.54	0.45
Asia-Asia	2,497.50	27.44	13.10	125.96	2,704.49	1.55	0.56
Afr-Afr	1,718.72	12.01	8.10	44.87	2,275.70	1.45	0.33

shows that the biggest vessels operate between Europe and Asia. The average distance of these routes is also the highest. In addition, on average the vessels on Europe-Asia are with 269 TEUs bigger than those on Asia-Europe routes.

Table 6 Average values according to the continents of origin and destination

Route	Mean TEU	Mean Ships	Mean Carriers	Mean Trips	Mean Distance	Mean Ships_per_carrier	Mean LSBCI
Eur-Asia	8,757.19	25.01	7.25	59.34	14,623.46	2.80	0.62
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Eur-Eur	4,241.52	54.07	23.23	328.18	2,306.26	2.04	0.66
Ame-Afr	4,206.42	11.60	5.65	30.13	10,785.53	1.49	0.44
Ame-Ame	3,912.69	23.15	11.14	87.91	2,794.57	1.78	0.50
Eur-Afr	3,343.37	23.76	11.20	98.68	3,423.56	1.75	0.46
Afr-Eur	2,969.62	18.10	10.46	87.06	3,495.96	1.54	0.45
Asia-Asia	2,497.50	27.44	13.10	125.96	2,704.49	1.55	0.56
Afr-Afr	1,718.72	12.01	8.10	44.87	2,275.70	1.45	0.33

A significant difference is observed between routes by number of trips. On Asia-Europe routes, the mean value of trips is 73 while on Europe-Asia ones it is 59 only, due the imbalance of trade between the two zones. A similar asymmetric pattern goes for the Asia-America route, with 67 trips and 24 ships on the front haul, against 56 trips and 21 ships on the back haul. However, the average size of companies measured in TEU per ship or in number of ships per company is very similar, unlike the Europe-Asia trade where, curiously, the average size of firms and vessels is slightly higher on the back haul (Europe to Asia) despite the imbalance. Almost symmetrical figures are reported on the routes between Europe and America. It indicates that the demand for container services are almost the same in the two continents. This trade balance facilitates carriers to operate efficiently, with an average firm size around 2 ships per carrier in both ways.

The intra-Asian and intra-African routes have different patterns. The average size of vessels as well as their absolute concentration measured by the number of ships per carrier is smaller than the intra-European and intra-American routes. The frequency of trips is fairly high on the intra-Asian trade routes compared to the intra-American trade, with 126 trips on average against 88, but both are far smaller than intra-European trade routes. Smaller vessels (1,719 TEU) operate on the intra-continental routes in Africa, where the bilateral connectivity between African countries is the lowest of the sample. Conversely, intra-Asian containerized trade is conducted between countries having a high connectivity. Appendix 1 depicts the maritime routes according to the region of origin and destination. Results show that the biggest vessels are deployed on the routes between East Asia, South-East Asia, Eastern Europe and Western Europe. On these routes, the average number of ships per vessels is the highest.

In summary, the overview of the 6,410 maritime show that there exist a significant misbalance between developed and less developed countries across regions. Maritime routes where larger companies compete (i.e. with a higher number of ships per carrier) are mainly associated with long distances and inter-regional trade between developed and less developed countries. Countries that are poorly connected tend to have a lower absolute concentration. In addition, intra-continental trade is more present between High Income countries and, as a consequence, their connectivity is also high. The market concentration (number of ships per carrier) is the highest on the routes connecting Developed (High income) countries and lowest on the links between “Less developed” ones. The trade routes “D-LD” and “LD-D” are not symmetrical, which indicates a misbalance within the industry. When we look at the continent of origin and destination, it seems that the highest disparities are on the routes between Europe and Asia, and America and Asia. This regional imbalance explains the strategies of shipping companies to deploy bigger ships and to decrease the trips on the back haul routes from Europe and America to Asia.

4.2. Principle Component Analysis (PCA)

Considering the heterogeneous nature of routes and the fact that competitors will focus their efforts on larger sub-markets, thus increasing competition, we have decided to restrict our analysis on the routes longer than 5,000 km. In order to include the routes with regular services, we have also removed from the sample those with less than 20 ships and 10 carriers. Finally, due to the presence of several outliers, routes with more than 1,000 trips were also removed. The final sample for the Principle Component Analysis consists of 800 maritime routes.

As active variables included in the PCA, we use Avg_TEU, Ships, Carriers, Trips, Distance, Ships_per_carrier and LSBCI. The supplementary variables are the following: Route_Country, Route_Continents, Route_Regions and Trade_Direction.

First we present the correlation coefficients of the active continuous and discrete variables.

Table 7 Pearson's correlations matrix for the subset of 800 maritime routes

	Avg_TEU	Ships	Carriers	Trips	Distance	Carrier_ships	LSBCI
Avg_TEU	1.000						
Ships	0.334	1.000					
Carriers	-0.079	0.789	1.000				
Trips	0.119	0.815	0.735	1.000			
Distance	0.481	0.141	-0.048	-0.043	1.000		
Ships_per_carrier	0.741	0.616	0.104	0.446	0.336	1.000	
LSBCI	0.475	0.282	0.037	0.152	0.180	0.480	1.000

We observe similar Pearson coefficients as in the correlation matrix of the whole sample of 6,410 routes (see **Error! Reference source not found.**). However, the correlation between the variables ships and carriers is lower in this subset of data. In addition, the trips and ships seem to have a stronger positive relation. The correlation coefficient of Avg_TEU and Ships_per_carrier as well as Avg_TEU and Carrier_ships are higher than those related to the previous dataset. When we conduct the Principle Component Analysis, all variables are standardized.

First we study the inertia of the principle components by analyzing the presence of correlation between variables and the components.

Table 8 Explained variance by component in the PCA analysis

Axis	Variance of the axis (eigenvalue)	% of explained variance	Cumulated % of explained variance	Retained axes according to Kaiser's criterion
1	3.196	45.7	45.7	X
2	1.957	28.0	73.6	X
3	0.817	11.7	85.3	
4	0.555	7.9	93.2	
5	0.261	3.7	96.9	
6	0.167	2.4	99.3	
7	0.047	0.7	100.0	
Total	7.000	100.0	100.0	

shows the Eigen value associated with each component, the percentage of explained variance and the cumulated percentage. Following the Kaiser's criterion, we have decided to retain the first two components. Their cumulative percentage of explained variance is 73.6%.

Next we evaluate the contribution of variables to each of the two components. As shown in **Error! Reference source not found.**, the average size of containerships, the distance between ports, ships per carrier and LSBCI have a significant and positive correlation with the first component and negative with the second one. On the other hand, the number of ships, carriers and trips correlate positively with both axes.

Table 9 Correlation coefficients of active variables with the axis and the contribution of the active variables to the axis

	Correlations between active variables and factors		Contributions of the active variables to the axes (in %)	
	Axis 1	Axis 2	Axis 1	Axis 2
Avg_TEU	0.617	-0.665	11.9	22.6
Ships	0.910	0.336	25.9	5.8
Carriers	0.601	0.702	11.3	25.2
Trips	0.767	0.521	18.4	13.9
Distance	0.332	-0.558	3.5	15.9

Ships_per_carrier	0.812	-0.386	20.6	7.6
LSBCI	0.518	-0.420	8.4	9.0

Active variables which have a major contribution in the construction of the first component are the number of ships and trips and, to a lesser extent, the average size of carriers. In other words, the most active routes are those where the traffic is intensive. Although linked to the first component, the weight of Distance is more limited compared to its influence on the second factor (negative correlation) where the number of carriers plays an opposite role (positive correlation). Therefore, the first component is related to the market size in terms of traffic and frequency of services, while the second component is associated with the increasing returns to scale of carriers facing long distance trade. Dropping the LSBCI variable and retaining the whole sample in the analysis -which is not presented here to avoid tedious presentation-, these two dimensions appear even more clearly: the first component captures the number of ships, trips and ships per operator, while the second axis introduces the negative correlation between the distance (associated with the average size of ships) and the number of carriers. Long distance hauls commands the absolute concentration of the market, with fewer and larger operating carriers.

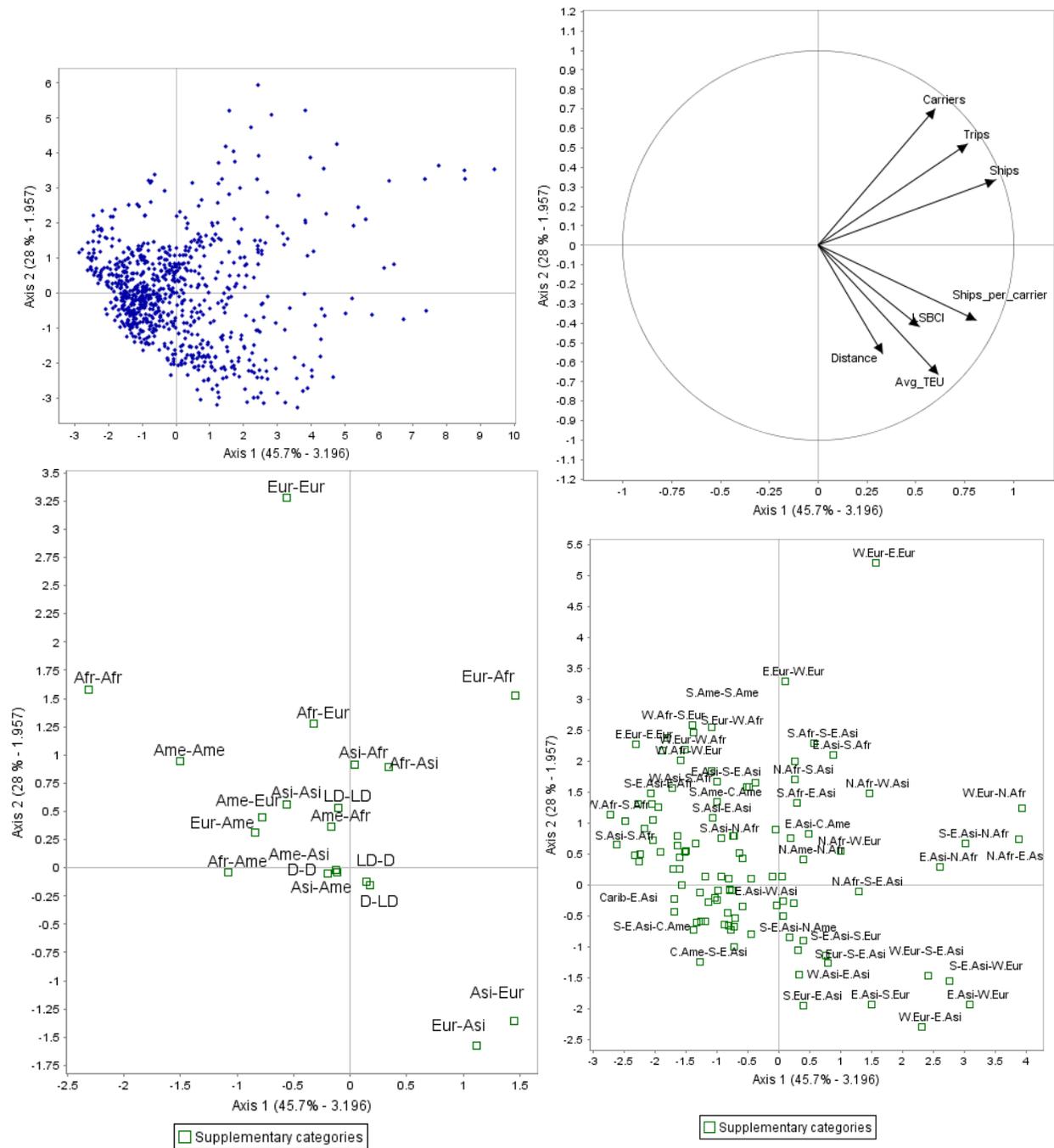


Figure 1: On the upper-left corner: the factorial map of individuals; On the upper- right corner: representation of the active continuous variables; On the lower-left corner: representation of the supplementary categorical variables related to the Trade Direction and the Trade between continents; On the lower-right corner: representation of the supplementary categorical variables related to trade between regions.

Looking at the illustrative variables related to the containerized trade between continents, we can see that within-Europe trade substantially contributes to the second axis. This route is

situated on the upper-left corner of the graph, which means that smaller vessels and a higher number of carriers operate on the intra-European trade routes. Intra-African trade is opposed to the routes connecting Europe and Asia. At the regional level, we can clearly see that the outliers in the sample are the routes connecting Western Europe and Eastern Europe. In addition, on the second axis are opposed Eastern Europe-Western Europe and South Europe-Eastern Asia routes. The former has a shorter sea distance, smaller vessels and more carriers than the latter. On the first axis, we can distinguish several groups based on the regions of origin and arrival. The routes connecting South-East or East Asia and Northern Africa can be characterized have more ships, trips and ships per carrier. On the other hand, routes between South Asia and South Africa, Caribbean and Central America, Caribbean and East-Asia, and South East Asia and Central America are perhaps more peripheral with a lower carrying capacity. Therefore, maritime routes have different characteristics and it seems useful to define different clusters based on the PCA active variables and described by both active and supplementary variables.

4.3. Cluster Analysis

The purpose of the Cluster Analysis is to classify a set of objects. The hierarchical method consists of agglomerating individuals which have similar or close values on active variables. This “cluster” is treated as one individual in a new matrix. This process continues until the optimal number of clusters is found and retained. To visualize these clusters, we build a hierarchical tree called dendrogram (see Figure 2). This tree can be considered a sequence of nested partitions from the most precise (in which each individual is a class), to the most general one (in which there is only one class where all inertia is lost) (Husson et al., 2017). As in previous section, we have used the same sample (800 maritime routes) to conduct the clustering analysis.

Table 10 Indicators of the quality of the identified clusters following Agglomerative Hierarchical Clustering

Criteria	3 clusters	4 clusters	5 clusters	6 clusters
Intra-cluster inertia	2.476	1.827	1.473	1.186
Inter-cluster inertia	2.676	3.325	3.679	3.966
Explained inertia (%)	51.937	64.543	71.414	76.974
Calinski-Harabasz (pseudo F) criterion	430.626	482.984	496.521	530.849
Davies-Bouldin's index	0.988	0.903	0.829	0.831

Error! Reference source not found. depicts that the optimal number of clusters is four. Even though, the Calinski-Harabasz criterion and Davies-Bouldin's index suggest that the optimum number of clusters are six and five respectively six and five, it should be highlighted that the inertia loss by forming six clusters is very small, less than 10%. Therefore, we have decided to retain four clusters. The quality of the clusters is relatively good due to the low intra-cluster and the high inter-cluster inertia (or variance). The former refers to the deviation between each point and the center of gravity of the cluster to which it belongs, while the latter is computed based on the deviation between each center of gravity for a specific cluster and the overall center of gravity.

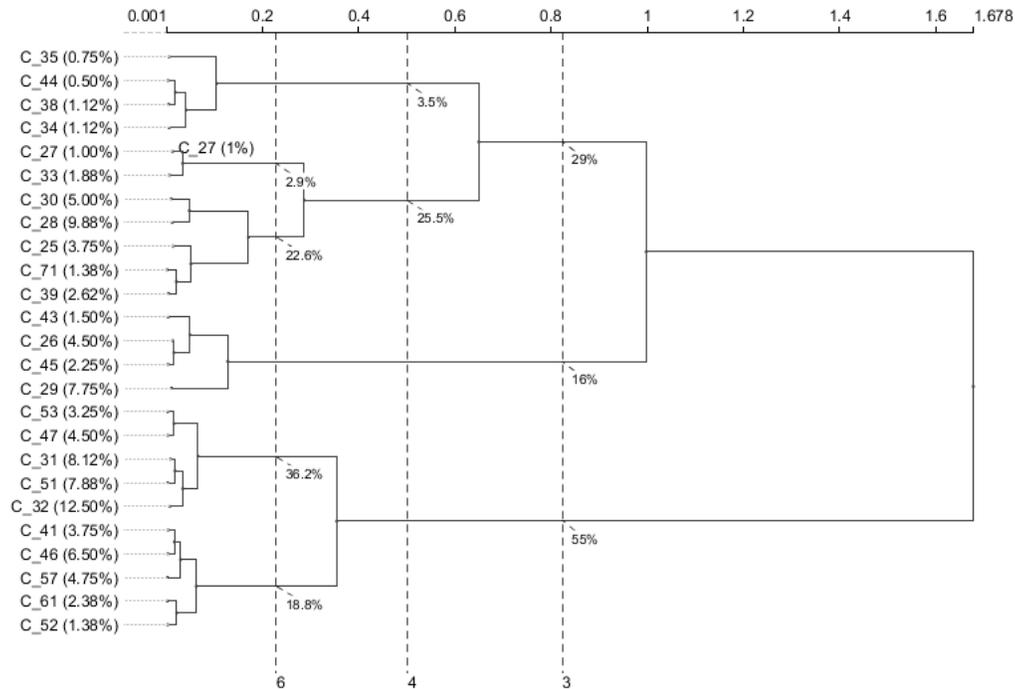


Figure 2 Dendrogram, Agglomerative Hierarchical Clustering based on the Ward criterion for the subset of 800 routes

Figure 2 shows the Hierarchical tree (Dendrogram) of the 800 maritime routes. By choosing to retain four clusters, we explain 64.5% of the total variability. These four clusters are depicted on the two axes that we have retained following the PCA (Figure 3). In the barycenter of cluster 1 is the maritime route connecting Tanjung Pelepas, Malaysia and Lagos, Nigeria. In 2014, between these two ports 36 ship with an average size of 4,293 TEU and 2.25 ships per carrier were operating. A typical representative of cluster two is the maritime connection between Tanjung Pelepas, Malaysia and Hamburg, Germany. There are 85 ships with a mean size of 9,850 TEU and 4.25 carriers per ships. Cluster three is represented by routes similar to the one between Shenzhen, China and Manzanillo, Mexico (89 ships, 6040 TEU and 2.07 ships per carrier). Finally, routes such as Ningbo-Zhoushan, China and Port Said, Egypt belong to the fourth cluster. On this route, there are 251 ships with an average size of 9,232 TEU and 4.45 ships per carrier.

transatlantic routes are over-represented in this first cluster. All mean variables are slightly lower than the mean values for the whole sample of 800 routes, the number of ships even representing nearly half the sample mean.

Table 11 Cluster 1 descriptive statistics of the active variables

Characteristic variables	Category mean	Overall mean	Category Std. deviation	Overall Std. deviation	Test-value	Probability
Distance	10,548.800	11,354.700	3,781.090	4,172.200	-6.400	0.000
LSBCI	0.556	0.597	0.105	0.109	-12.463	0.000
Carriers	17.879	22.656	5.418	11.588	-13.659	0.000
Avg_TEU	5,381.440	6,529.610	1,602.970	2,552.900	-14.901	0.000
Trips	111.095	174.604	60.626	137.482	-15.305	0.000
Ships_per_carrier	2.109	2.619	0.442	0.994	-17.008	0.000
Ships	37.043	60.546	11.669	45.545	-17.097	0.000

Table 12 Cluster 1, descriptive statistics of the supplementary variables

Variable label	Characteristic categories	% of category in group	% of category in set	% of group in category	Probability
Trade_direction	D-D	37.1	33.6	63.9	0.008
Trade_direction	D-LD	21.8	25.6	49.3	0.003
Route_Continents	Eur-Ame	14.7	10.4	81.9	0.000
Route_Continents	Ame-Eur	11.0	7.9	81.0	0.000
Route-regions	S.Eur-N.Ame	5.6	3.5	92.9	0.000
Route_Continents	Ame-Ame	5.6	3.8	86.7	0.001

The absolute concentration within this cluster is relatively low, with 2.1 carriers per ship on average, a vessel size below average and only 18 carriers against 23 per route in the selected sample.

4.3.2. Cluster 2, low degree of competition (170 maritime routes)

Some 170 maritime routes are included in the second cluster. All active variables except the number of trips are significant. This group is mainly linked with the second component (Table 14 *Cluster 2, descriptive statistics of the supplementary variables*). The routes are significantly longer

(14,730 nautical miles) and characterized by fewer carriers operating nearly 4 ships on average and deploying rather large vessels (Avg_TEU is 10,149 TEU).

Table 13 Cluster 2, descriptive statistics of the active variables

Characteristic variables	Category mean	Overall mean	Category Std. deviation	Overall Std. deviation	Probability
Avg_TEU	10,148.800	6,529.610	1,468.070	2,552.900	0.000
Ships_per_carrier	3.764	2.619	0.946	0.994	0.000
LSBCI	0.687	0.597	0.058	0.109	0.000
Distance	14,729.900	11,354.700	3,821.480	4,172.200	0.000
Ships	69.624	60.546	26.788	45.545	0.002
Trips	170.953	174.604	80.117	137.482	0.348
Carriers	18.559	22.656	5.634	11.588	0.000

Table 14 Cluster 2, descriptive statistics of the supplementary variables

Variable label	Characteristic categories	% of category in group	% of category in set	% of group in category	Probability
Route_Continents	Eur-Asi	41.2	11.1	78.7	0.000
Trade_direction	D-L	35.3	25.6	29.3	0.001
Route_Continents	Asi-Eur	34.1	11.0	65.9	0.000
Route-regions	W.Eur-E.Asi	13.5	2.9	100.0	0.000
Route-regions	S.Eur-E.Asi	11.8	2.5	100.0	0.000
Route-regions	E.Asi-W.Eur	10.0	2.8	77.3	0.000
Route-regions	S-E.Asi-W.Eur	8.2	2.0	87.5	0.000
Route_Continents	Ame-Asi	6.5	14.8	9.3	0.000
Trade_direction	L-L	6.5	16.3	8.5	0.000

From a regional perspective, this cluster gathers nearly all front and back hauls between Europe and Asia which represent three quarters of the cluster. The ports of departure or arrival for these lines have greater connectivity (nearly 10 points higher) than the rest of the sample. Absolute concentration of this sub-market is undoubtedly higher than previous cluster.

4.3.3. Cluster 3, high degree of competition (135 maritime routes)

There are 135 routes in the third cluster, mainly between Asia and America. The trade between China and USA is substantial in this group. Unlike the second cluster, this group is mainly linked

with the first component (market size). In this cluster, there are considerably more ships and trips than average. The variables which characterized the cluster are therefore Carriers, Trips and Ships. The average distance in this group is the shortest, with 9,549.44 km. On the other hand, competition on the routes is relatively intense between smaller carriers (2.46 ships per carrier, against 2.62 in the sample). The average size of vessels (around 5,300 TEU) is comparable to that of Cluster 1.

Table 15 Cluster 3, descriptive statistics of the active variables

Characteristic variables	Category mean	Overall mean	Category Std. deviation	Overall Std. deviation	Probability
Carriers	37.630	22.656	9.900	11.588	0.000
Trips	304.052	174.604	140.208	137.482	0.000
Ships	91.437	60.546	26.711	45.545	0.000
LSBCI	0.608	0.597	0.100	0.109	0.102
ships_per_carrier	2.464	2.619	0.513	0.994	0.023
Distance	9,549.440	11,354.700	3,245.700	4,172.200	0.000
Avg_TEU	5,336.050	6,529.610	1,530.720	2,552.900	0.000

Table 16 Cluster 3, descriptive statistics of the supplementary variables

Variable label	Characteristic categories	% of category in group	% of category in set	% of group in category	Probability
Route_Continents	Asi-Ame	20.7	14.1	24.8	0.013
Route_Continents	Ame-Asi	20.7	14.8	23.7	0.025
Route-regions	N.Ame-E.Asi	15.6	8.5	30.9	0.002
Route_Country	USA-CHN	11.1	3.8	50.0	0.000
Route_Country	CHN-USA	6.7	3.3	34.6	0.020
Route-regions	E.Asi-C.Ame	5.2	2.0	43.8	0.010

4.3.4. Cluster 4, low degree of competition (32 maritime routes)

The last one is the smallest cluster, with only 32 maritime routes. This cluster is mainly linked with the first axis, like previous cluster, and the most significant variables describing it are also the number of ships, trips, and carriers. In this cluster are found the trade lines operated either between developed or between less developed countries located mainly in Eastern Asia and Western Europe.

Table 17 Cluster 4, descriptive statistics of the active variables

Characteristic variables	Category mean	Overall mean	Category Std. deviation	Overall Std. deviation	Probability
Ships	222.063	60.546	67.258	45.545	0.000
Trips	566.781	174.604	139.269	137.482	0.000
Carriers	50.375	22.656	14.711	11.588	0.000
Ships_per_carrier	4.581	2.619	1.161	0.994	0.000
Avg_TEU	8950.870	6529.610	1972.240	2552.900	0.000
LSBCI	0.666	0.597	0.064	0.109	0.000
Distance	12,701.600	11354.700	4563.120	4172.200	0.031

Table 18 Cluster 4, descriptive statistics of the supplementary variables

Variable label	Characteristic categories	% of category in group	% of category in set	% of group in category	Probability
Trade_direction	LD-LD	37.5	16.3	9.2	0.003
Route_Continents	Afr-Asi	25.0	5.8	17.4	0.000
Route-regions	E.Asi-W.Eur	15.6	2.8	22.7	0.001
Route_Continents	Asi-Afr	15.6	5.4	11.6	0.024
Trade_direction	D-D	12.5	33.6	1.5	0.006

The degree of competition on the routes in the last cluster is the lowest: on average, there are 4.58 ships per carrier. Surprisingly, these are the busiest routes, the mean values of the number of ships and trips are considerably higher than average, 266% and 244% respectively. As shown in Figure 3, most of the routes are between hub ports such as Port Said in Egypt situated near the

Suez Canal, the Port of Singapore -the biggest port in ASEAN-, the port of Rotterdam -biggest port in Europe- as well as the port of Shanghai, -biggest port in the world. This is confirmed by the higher connectivity degree (0.67), close to that observed in the second cluster. These direct connections between hub-ports seem to limit the degree of competition since, they are dominated mainly by large companies operating large vessels.

4.3.5. Discussion of the typology and absolute concentration

In summary, the cluster analysis allows us to identify four categories of maritime routes based on the active variables. We observe that the lowest degree of competition (4.58 ships per carriers) is within a small clusters of 32 very busy routes connecting large hub-ports. The second cluster also is constituted of routes with fewer large carriers connecting mainly Western Europe with Eastern and South-Eastern Asia. The degree of competition is the highest among the routes between developed countries (intra-American and Intra-European routes). The container flows on these routes are relatively symetrical which do not require a high degree of cooperation between carriers to optimize their costs (slot capacity). The degree of competition is also relatively high on the routes connecting America and Asia where substantial part of the trade is conducted between China and USA.

An Interactive Decision Tree (IDT) based on the CART approach (Breiman et al. 1984) scrutinizes the typology by highlighting the most discriminating continuous variables of the absolute concentration of maritime routes measured by the average size of vessels (Fig. 4).

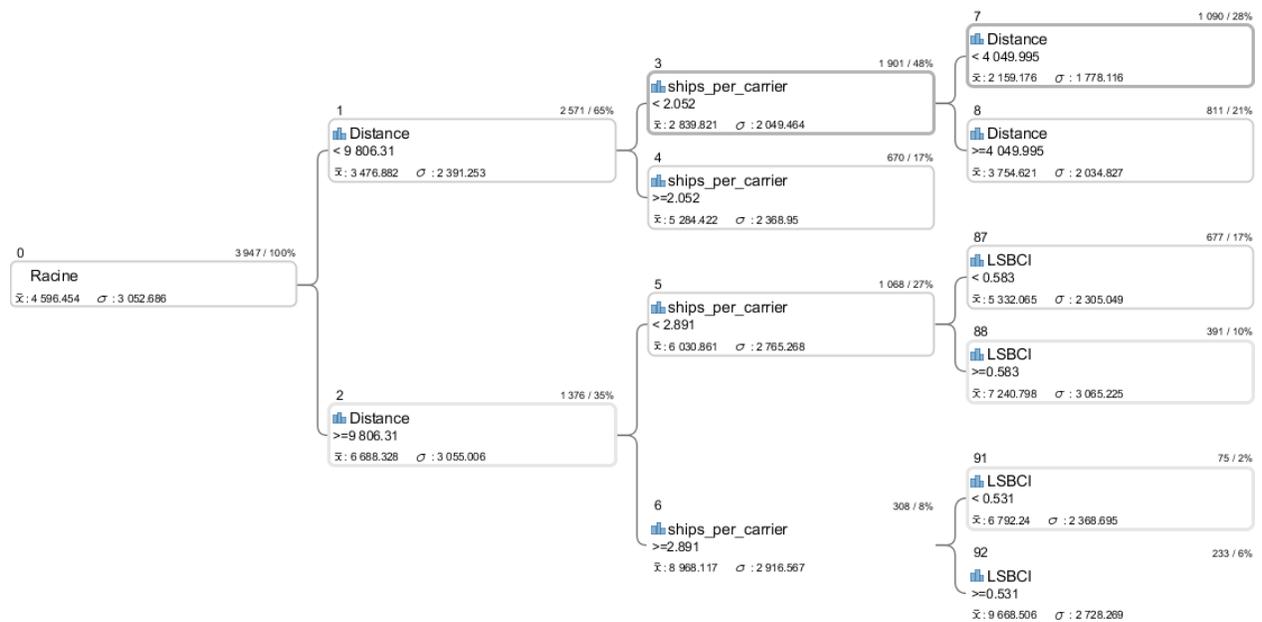


Figure 4. Interactive Decision Tree of the average size of vessels in TEU

(Legend: the darker the box at the end of a branch, the smaller the size relatively to the mean of the sample)

Starting with an average size of 4,596 TEU per ship (st.-dev. 3,053 TEU), the first criterion to split up the sample is the distance of the route in nautical miles, below or above 9,806 nautical miles. Then comes the size of companies proxied by the number of ships per carrier. For shorter routes, the threshold is below or above 2.05 ships per carrier. For longer routes, this threshold is below or above 2.9. In other words, the companies owning less than 2.05 ships on routes shorter than 9,806 miles will operate the smallest containerships (2,839 TEU). By contrast, those carriers owning more than 2.9 ships on routes longer than 9,806 miles will operate large containerships of 8,968 TEU on average. The last discriminating criterion to increase further the average size of vessels is the connectivity of ports ($LSBCI > 0.53$): operating vessels between hub ports is likely to increase the size nearly up to 10,000 TEU.

5. Conclusions

The degree of competition on the maritime routes was evaluated through the simplest concentration ratios available in the Lloyd's List database, which are the number of ships per carrier and the average size of containerships deployed on the route. Unfortunately, unlike other studies focusing on a more limited number of routes, we did not have access to the distribution of market shares between liner companies for every route. Firstly, a descriptive statistics analysis of a sample of 6,410 routes between 153 ports reveals a regional misbalance between ports of origin and destination, particularly between Asia and Europe, and Asia and America. The biggest vessels are deployed on these routes because of the long distance and the high degree of port connectivity with other ports, especially between Europe and Asia. These routes are also the most concentrated. The highest levels of competition measured by these simple instruments is observed on the intra-African routes, which can be characterized as the shortest distance between ports, smallest vessels, the lowest number of trips and the worst bilateral country connectivity. Furthermore, when we look at the origin and destination across regions, we conclude that the highest level of absolute concentration of the market is between

Eastern Asia and Eastern Europe, South-East Asia and Eastern Europe, Southern Europe and Western Asia, and Western Europe and South-East Asia. It was found that competition is stiffer on the routes connecting mainly less developed regions such as Northern Africa and Eastern Asia.

Considering the heterogeneous nature of the whole sample we have decided to focus on the longer distance routes where are conducted many trips and multiple container vessels operate. Therefore, the sample size was limited to 800 routes. By applying Principle Component Analysis, we have identified two components which summarize the seven active variables and explain 73.6% of the variance. The first component rather relates to the market size, frequency of services and competition (number of ships, trips, carriers) while the second rather illustrates a certain form of absolute concentration (average size of ships) positively linked with the length of the route. For this limited sample, the two components are not so much independent one from the other, unlike the same analysis conducted over the whole sample of maritime routes where these two dimensions are less linked together.

The PCA-based cluster Analysis has allowed us to identify 4 distinct clusters. The first one includes mainly transatlantic routes. These are the routes where competition is less intense (lower number of ships, trips, carriers than average) which can be explained by low level of regional misbalance, which allows companies to operate efficiently and probably without having to share slot capacity with their competitors. The third cluster is similar to the first one and includes mainly routes between Asia and America. It is slightly more concentrated even though the distance and the average size of vessels are almost the same.

On the other hand, the second cluster is made up mainly with routes between Europe and Asia. The absolute market concentration is higher in this cluster, 3.76 ships per carrier. The distance and size of vessels have the highest mean values among the clusters. The group with the lowest degree of competition is in the fourth cluster, which has the largest hub ports in the network and despite the lower mean values of distance and the size of the ships than the second cluster, the level of market concentration is the highest. These results suggest that the distance, frequency of services and the size of the ships do have an impact on the degree of competition.

However, between major hub-routes were registered the highest frequency of services and at the same time the highest degree of monopolization. This is confirmed by the interactive regression tree explaining the average size of vessels: the criterion of distance comes first to explain this variable, followed by the size of the company and the connectivity index.

In Summary, we have shown that large distance routes between connected ports are in general more concentrated in absolute terms (fewer companies operating larger vessels), exploiting the large economies of scale on these routes. However, the degree of competition varies across regions of origin and destination. The competition looks more intense on intra-African and intra-Asian routes and also between less developed countries. The routes where the degree of competition is also relatively high is intra-continental trade in Europe and America. On the other hand, the trade misbalance between Asia and America and Asia and Europe might be one of the main reason for the relatively higher concentration on these routes. The PCA has shown that the Hub-and-Spoke network of the maritime transport leads to a smaller number of shipping companies operating a larger fleet of bigger vessels between the largest ports and these routes have the highest degree of concentration.

Additional studies should be conducted to analyze the nature of competition exclusively between the largest hub-ports in the maritime network or between smaller (feeder) ports. Additional data such as, the distribution of market shares per carrier on specific routes, the exact number of containers transported by each vessels from the port of origin to the destination as well as the amount of transshipments by vessel would allow researchers to understand better the nature of competition within the containerized trade network.

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Appendices

Appendix 1: Regional container trade between the 153 ports in the sample.

	Mean TEU	Mean Ships	Mean Carriers	Mean Trips	Mean Distance	Mean Ships_per_carrier	Mean LSBCI
E.Asi-E.Eur	15377	15.00	3.00	62.00	14363	5.00	0.533
E.Eur-E.Asi	13423	9.00	3.00	21.00	21543	3.00	0.537
S-E.Asi-E.Eur	12166	7.00	1.67	31.33	13645	3.00	0.503
E.Eur-S-E.Asi	10889	10.00	2.33	30.33	13520	3.56	0.468
W.Eur-E.Asi	10003	26.08	6.94	55.86	19624	2.66	0.676
E.Asi-W.Eur	9867	32.33	8.07	79.29	18183	2.90	0.682
W.Eur-S-E.Asi	9524	48.71	10.65	135.10	15161	3.91	0.627
E.Asi-S.Eur	9173	24.12	6.75	57.12	14289	2.67	0.630
S-E.Asi-W.Eur	9036	56.34	12.31	172.28	15311	3.62	0.639
S-E.Asi-S.Eur	8900	36.82	11.57	101.61	11573	2.79	0.611
E.Asi-N.Afr	8897	28.00	8.50	58.62	13440	1.69	0.484
N.Afr-E.Asi	8714	35.37	8.76	68.57	15427	1.73	0.456
W.Eur-W.Asi	8669	30.08	9.36	63.19	9259	3.03	0.591
S.Eur-E.Asi	8337	15.96	5.65	31.91	16184	2.31	0.633
S-E.Asi-N.Afr	8146	52.07	13.27	132.07	11374	2.12	0.444
S.Ame-S-E.Asi	7820	30.88	12.88	93.25	17892	1.71	0.473
E.Asi-W.Asi	7806	14.16	6.69	24.82	9889	1.77	0.574

S.Ame-S.Afr	7736	51.50	27.50	152.50	8195	1.87	0.450
N.Afr-S.Ame	7643	30.50	14.00	138.00	10240	2.17	0.477
S.Eur-W.Asi	7642	25.12	7.24	67.12	6097	3.94	0.577
S.Eur-S-E.Asi	7584	29.29	9.18	82.07	11561	2.61	0.572
W.Asi-E.Asi	7500	17.37	6.70	26.19	10676	1.89	0.604
S-E.Asi-S.Ame	7368	31.11	12.33	114.44	18065	1.93	0.466
W.Eur-S.Asi	7281	18.42	7.74	48.68	11910	2.19	0.553
S.Ame-S.Asi	7186	8.00	4.00	12.67	15385	1.90	0.399
C.Ame-W.Asi	7172	1.33	1.33	1.67	19255	1.00	0.418
S.Afr-S.Ame	7081	30.50	17.50	94.50	8195	1.74	0.450
S.Ame-N.Afr	6845	15.75	7.25	86.25	12140	1.86	0.447
S.Asi-W.Eur	6665	18.71	7.79	45.54	12018	2.19	0.553
S.Afr-W.Eur	6656	12.43	5.29	59.57	12830	3.04	0.519
E.Asi-S.Ame	6637	34.17	14.00	87.57	16245	2.37	0.484
W.Eur-S.Afr	6637	14.00	6.29	58.86	12831	2.78	0.519
W.Asi-W.Eur	6611	14.68	5.90	34.45	9218	2.49	0.586
S.Ame-E.Asi	6609	20.62	9.08	43.50	18768	1.90	0.486
W.Asi-Carib	6575	6.50	4.50	8.50	15880	1.69	0.339
Carib-W.Asi	6509	8.00	4.50	8.50	15880	3.19	0.339
S.Ame-W.Eur	6288	18.90	9.29	89.48	10796	1.90	0.479
N.Afr-S-E.Asi	6220	44.86	10.95	124.95	11206	1.84	0.406
W.Asi-C.Ame	6218	2.40	1.60	2.60	18124	1.50	0.415
S.Eur-S.Asi	6119	8.59	3.35	27.65	8521	2.14	0.525
N.Ame-S-E.Asi	6066	11.30	4.14	29.02	13613	2.25	0.529
W.Eur-S.Ame	5997	15.80	8.04	74.44	10837	1.74	0.478
S.Asi-S.Eur	5996	10.83	4.54	33.00	8570	2.18	0.520
Carib-S.Asi	5936	5.50	2.50	5.75	17715	3.36	0.371
S-E.Asi-N.Ame	5887	12.80	4.92	35.18	13726	2.11	0.555
N.Afr-W.Asi	5847	34.58	11.89	80.58	5904	1.75	0.416
S.Ame-W.Asi	5803	1.50	1.50	2.00	15706	1.00	0.398
E.Asi-N.Ame	5701	28.36	11.72	83.95	8903	2.11	0.682
Carib-E.Asi	5666	11.31	6.92	22.46	17531	1.63	0.451
N.Ame-E.Asi	5648	27.22	11.21	79.14	10025	2.05	0.683
W.Asi-N.Ame	5600	11.09	5.91	24.75	16982	1.77	0.543
N.Ame-S.Asi	5580	10.67	4.63	23.25	17515	2.22	0.536
W.Asi-S-E.Asi	5564	28.00	10.19	41.89	6854	2.03	0.514
E.Asi-Carib	5532	17.33	10.08	48.17	14885	1.65	0.459
S-E.Asi-W.Asi	5470	29.36	12.64	43.50	6675	1.85	0.507
N.Afr-N.Ame	5437	10.47	5.42	21.78	10009	1.51	0.524
N.Afr-S.Asi	5418	23.67	10.33	52.56	7793	1.59	0.439
N.Ame-W.Asi	5234	12.26	5.80	28.49	16795	2.14	0.547

S.Eur-S.Ame	5205	10.47	5.16	74.26	10988	2.05	0.448
W.Asi-S.Eur	5185	20.89	8.49	55.94	5881	2.40	0.571
E.Asi-C.Ame	5145	52.09	21.47	127.74	13301	1.97	0.508
W.Asi-N.Afr	5129	21.50	9.94	58.39	5528	1.59	0.429
S.Asi-C.Ame	5122	1.20	1.20	1.20	19319	1.00	0.403
Carib-S-E.Asi	5117	4.00	1.40	4.80	19807	3.60	0.380
C.Ame-S.Asi	5060	1.00	1.00	1.00	18907	1.00	0.422
C.Ame-S.Afr	5059	1.00	1.00	1.00	13225	1.00	0.430
S.Eur-E.Eur	5046	6.80	2.80	21.00	3700	1.85	0.469
E.Eur-N.Afr	5027	11.75	7.88	43.63	4186	2.05	0.477
W.Eur-E.Afr	5026	13.14	3.29	38.71	7625	4.02	0.357
N.Afr-Carib	5007	1.33	1.33	1.33	8425	1.00	0.331
S.Asi-N.Afr	5001	15.17	7.25	34.00	8094	1.64	0.413
S.Eur-S.Afr	4984	6.71	3.29	31.71	11218	2.67	0.466
S-E.Asi-Carib	4967	5.00	1.75	6.00	20057	2.08	0.416
W.Eur-W.Eur	4900	127.20	49.12	777.82	1393	2.40	0.821
S.Afr-S.Eur	4899	7.14	4.57	34.71	11218	1.92	0.466
N.Ame-N.Afr	4868	14.47	6.37	30.81	9778	1.53	0.490
S.Ame-S.Eur	4848	9.10	4.05	60.90	10750	2.07	0.448
S.Asi-N.Ame	4813	11.72	6.06	28.17	17419	1.65	0.533
W.Eur-S.Eur	4810	25.43	11.22	102.94	2724	2.10	0.675
Carib-S.Eur	4804	12.25	6.75	22.08	7367	1.71	0.424
C.Ame-S-E.Asi	4782	7.78	5.22	9.89	17402	1.22	0.413
S.Afr-S.Asi	4766	24.50	10.00	48.00	6935	2.45	0.411
C.Ame-E.Asi	4747	39.82	17.85	89.76	14658	1.82	0.511
S.Asi-Carib	4672	3.25	2.25	4.00	17606	1.17	0.368
S.Eur-W.Eur	4666	22.40	10.50	102.76	2738	1.90	0.676
E.Afr-W.Eur	4658	6.25	2.50	12.63	7871	2.26	0.357
E.Afr-N.Ame	4630	4.33	3.33	12.33	15530	1.13	0.396
Carib-W.Eur	4628	12.25	7.17	45.67	6546	1.78	0.429
C.Ame-S.Ame	4616	39.25	17.67	157.25	7381	2.11	0.439
W.Eur-N.Ame	4597	19.53	8.73	68.81	8080	1.97	0.661
W.Asi-W.Asi	4553	58.91	24.09	211.45	2734	2.28	0.581
S.Asi-E.Asi	4544	8.76	5.27	19.15	8248	1.42	0.534
W.Eur-Carib	4496	14.38	7.69	52.08	6537	1.88	0.427
S-E.Asi-C.Ame	4475	9.00	5.89	12.78	17338	1.19	0.440
N.Ame-W.Eur	4461	18.46	7.93	63.26	7880	2.16	0.651
S.Eur-Carib	4450	9.27	5.67	19.07	7482	1.39	0.416
S.Ame-Carib	4440	16.00	8.00	53.13	6919	2.87	0.380
W.Eur-N.Afr	4428	31.42	12.16	89.82	2606	1.62	0.498
S.Eur-N.Ame	4421	15.32	6.80	46.87	8753	2.28	0.646

N.Ame-S.Eur	4395	13.31	6.04	40.20	8888	2.17	0.637
E.Eur-S.Eur	4392	3.57	2.43	5.14	3354	1.39	0.456
E.Asi-E.Afr	4388	10.20	5.47	38.60	9468	1.46	0.380
Carib-S.Ame	4379	16.25	8.13	52.25	6920	2.50	0.380
C.Ame-E.Afr	4375	1.50	1.50	2.00	16525	1.00	0.296
N.Afr-E.Afr	4289	20.00	7.67	52.33	4747	3.41	0.355
N.Ame-S.Ame	4264	14.47	8.91	54.95	8349	1.51	0.488
E.Afr-N.Afr	4249	15.33	8.00	34.67	4747	2.47	0.355
E.Afr-E.Asi	4225	11.13	5.20	33.60	10939	1.57	0.373
E.Asi-S.Asi	4206	6.73	3.71	14.63	7940	1.31	0.534
N.Ame-E.Afr	4204	10.25	7.00	30.25	15532	1.36	0.396
S.Afr-E.Asi	4194	29.31	14.81	59.50	13018	1.50	0.515
Carib-Carib	4150	22.00	7.50	53.00	1156	2.95	0.386
S.Ame-N.Ame	4136	14.36	8.90	54.67	7963	1.52	0.494
C.Ame-S.Eur	4101	15.52	7.84	50.84	9837	2.18	0.455
E.Asi-S.Afr	4054	34.69	17.31	72.69	12087	1.47	0.531
W.Afr-N.Ame	3987	1.00	1.00	1.00	9614	1.00	0.384
S.Ame-S.Ame	3970	54.50	24.10	253.30	6187	1.82	0.411
N.Ame-C.Ame	3968	22.95	11.80	79.36	2908	1.59	0.529
S.Eur-C.Ame	3958	14.63	7.44	49.59	9735	2.03	0.445
N.Ame-N.Ame	3944	24.30	10.70	91.86	253	1.85	0.622
Carib-C.Ame	3904	25.29	14.14	92.29	2941	1.79	0.381
S.Afr-S-E.Asi	3868	48.83	24.50	120.50	8944	1.45	0.459
N.Afr-C.Ame	3828	5.57	3.71	12.57	10155	1.28	0.373
Carib-N.Ame	3812	19.14	9.57	84.50	1889	1.95	0.517
S.Ame-C.Ame	3758	28.08	15.54	130.38	7408	1.41	0.417
W.Eur-C.Ame	3753	26.04	12.89	105.93	8778	1.64	0.466
S.Asi-S.Afr	3720	24.00	10.00	45.50	6935	2.52	0.411
S.Asi-W.Asi	3714	22.77	11.23	45.15	2793	1.79	0.536
E.Eur-S.Ame	3706	1.00	1.00	1.00	14481	1.00	0.390
S-E.Asi-W.Afr	3698	38.71	22.86	81.43	16030	1.64	0.403
C.Ame-Carib	3653	23.13	13.38	81.88	3217	1.77	0.380
N.Afr-W.Eur	3641	19.54	10.54	62.92	2687	1.46	0.503
S.Eur-S.Eur	3566	62.63	26.52	352.87	1162	2.18	0.620
N.Ame-Carib	3555	24.79	11.04	96.50	1868	2.27	0.509
W.Asi-S.Asi	3535	26.83	13.00	45.17	2739	1.52	0.539
W.Afr-E.Asi	3520	13.82	9.18	19.55	20247	1.46	0.425
E.Afr-S.Asi	3459	13.80	9.40	38.80	3781	1.56	0.370
C.Ame-N.Ame	3457	23.45	11.75	79.79	2882	1.53	0.534
C.Ame-W.Eur	3453	25.15	12.50	106.19	8543	1.65	0.461
E.Afr-S.Eur	3448	6.57	3.14	18.14	6101	2.53	0.372

Carib-N.Afr	3448	4.75	2.75	5.25	9102	1.40	0.346
Carib-E.Eur	3436	6.00	4.50	24.50	10816	1.19	0.373
E.Asi-W.Afr	3316	24.08	14.08	40.17	16809	1.48	0.421
E.Afr-S-E.Asi	3295	15.64	9.71	49.36	7491	1.37	0.349
N.Afr-E.Eur	3280	7.67	6.67	22.78	4118	1.16	0.429
S-E.Asi-E.Afr	3280	16.42	11.58	55.42	7343	1.33	0.350
S-E.Asi-S.Afr	3165	43.67	21.00	117.50	8944	1.53	0.459
S.Afr-W.Asi	3161	29.25	16.75	74.50	7476	1.69	0.414
N.Ame-W.Afr	3128	1.00	1.00	1.00	10407	1.00	0.396
W.Afr-E.Eur	3113	15.00	5.00	30.00	8916	3.00	0.324
W.Asi-S.Afr	3112	28.25	16.00	76.00	7476	2.27	0.414
N.Afr-S.Afr	3060	2.67	1.67	5.67	10873	1.33	0.361
E.Afr-W.Asi	3040	30.29	16.14	115.00	3079	1.91	0.382
E.Eur-W.Afr	3035	9.00	2.00	14.00	8916	4.50	0.324
W.Afr-S-E.Asi	2952	19.00	12.00	34.15	16842	1.60	0.344
S-E.Asi-S.Asi	2792	18.41	10.97	43.15	4405	1.35	0.501
W.Afr-S.Afr	2776	26.00	16.00	39.00	6950	1.64	0.384
C.Ame-C.Ame	2767	28.20	14.10	108.70	2160	1.35	0.440
S.Asi-S-E.Asi	2764	14.21	8.82	32.42	4531	1.23	0.485
S.Afr-N.Ame	2764	8.29	4.14	33.14	14846	1.69	0.470
S.Asi-E.Afr	2761	16.25	9.50	45.50	3951	1.58	0.367
S.Eur-E.Afr	2757	5.36	2.64	18.09	5990	2.23	0.369
W.Asi-E.Afr	2738	24.38	12.50	99.87	3069	1.98	0.378
W.Afr-C.Ame	2732	1.00	1.00	1.00	8949	1.00	0.328
W.Eur-W.Afr	2724	11.77	8.00	43.38	7027	1.48	0.424
W.Afr-W.Afr	2695	58.00	37.00	126.50	819	1.51	0.422
S-E.Asi-E.Asi	2640	20.85	10.88	65.54	3923	1.43	0.554
C.Ame-N.Afr	2593	4.70	3.50	12.30	10348	1.24	0.370
C.Ame-E.Eur	2568	9.50	5.67	29.50	10656	1.59	0.351
Carib-S.Afr	2558	11.00	4.00	19.00	13353	2.75	0.359
W.Afr-W.Eur	2558	11.54	7.23	43.92	7046	1.49	0.420
E.Eur-C.Ame	2557	10.17	6.33	31.33	10656	1.53	0.351
E.Asi-S-E.Asi	2550	20.18	10.55	64.70	3592	1.39	0.551
S.Eur-W.Afr	2544	20.78	13.78	69.00	6258	1.36	0.421
E.Afr-E.Afr	2533	10.00	5.00	18.50	3223	2.00	0.316
W.Afr-S.Eur	2506	16.08	11.69	51.38	6146	1.22	0.425
S.Afr-E.Afr	2454	9.50	5.50	18.50	4925	1.68	0.343
S.Afr-Carib	2452	10.00	3.00	21.00	13354	3.33	0.359
E.Eur-Carib	2425	5.50	4.00	24.50	9514	1.21	0.348
W.Afr-S.Asi	2424	8.25	4.75	9.25	14311	3.12	0.365
S.Afr-W.Afr	2353	23.00	12.00	44.00	6949	1.89	0.384

E.Eur-W.Eur	2352	41.44	26.25	492.19	3802	1.41	0.476
W.Afr-N.Afr	2266	19.83	13.17	65.83	5885	1.55	0.407
S.Afr-N.Afr	2257	2.33	1.67	2.67	10701	1.67	0.428
E.Eur-N.Ame	2225	2.80	2.00	8.40	8409	1.57	0.518
E.Afr-S.Afr	2147	12.50	4.00	27.50	4925	3.08	0.343
N.Afr-W.Afr	2074	21.60	14.40	76.80	5418	1.59	0.379
N.Ame-E.Eur	2063	2.60	1.60	9.00	8410	1.70	0.518
N.Ame-S.Afr	2057	6.88	3.25	31.00	14869	1.70	0.472
S.Eur-N.Afr	2034	26.30	13.87	148.16	1005	1.47	0.441
E.Asi-E.Asi	1981	46.34	20.22	239.92	596	1.63	0.694
S.Asi-W.Afr	1969	10.50	4.00	15.50	14237	6.21	0.374
S.Asi-S.Asi	1946	8.50	5.65	19.50	722	1.16	0.561
W.Afr-W.Asi	1924	3.50	3.50	3.50	14930	1.00	0.365
N.Afr-S.Eur	1898	22.86	13.47	145.84	1036	1.40	0.432
W.Eur-E.Eur	1885	41.00	26.44	484.94	3960	1.32	0.476
S.Ame-W.Afr	1848	2.33	2.33	5.67	6384	1.00	0.328
W.Afr-S.Ame	1706	3.00	3.00	5.00	6556	1.00	0.331
W.Asi-W.Afr	1686	8.75	2.75	14.00	14928	6.46	0.365
E.Afr-W.Afr	1598	8.00	1.00	9.00	10670	8.00	0.273
E.Eur-E.Eur	1583	30.00	22.50	218.50	7845	1.33	0.338
N.Afr-N.Afr	1377	10.05	7.36	44.81	917	1.23	0.313
S-E.Asi-S-E.Asi	1286	17.38	9.76	114.63	1160	1.59	0.430
E.Eur-S.Afr	259	2.00	1.00	5.00	11544	2.00	0.403
S.Afr-E.Eur	142	1.00	1.00	8.00	11545	1.00	0.403