

Appendix IV Global Innovation Index science and technology cluster methodology

Since 2016 the Global Innovation Index (GII) has sought to identify Science and Technology (S&T) clusters using a bottom-up approach. This approach disregards administrative or political borders and instead pinpoints those geographical areas showing a high density of inventors and scientific authors. The resultant clusters often encompass several municipal districts, sub-federal states, and sometimes even two or more countries.

The same methodology used in previous editions of the GII was employed in the compilation of this year's list of the top 100 GII S&T clusters worldwide (Bergquist and Fink, 2020: 43–63). It comprised:

- selecting inventors listed in published patent applications under WIPO's Patent Cooperation Treaty (PCT) spanning the period 2015 to 2019;
- selecting authors listed in scientific publications in the Web of Science's Science Citation Index Expanded (SCIE) covering the same period;
- geocoding inventor and author addresses and then applying the density-based spatial clustering of applications with noise (DBSCAN) algorithm to the geocoded inventor and author points.

The WIPO PCT patent dataset consists of approximately 1.1 million patent applications published between 2015 and 2019 containing 3.2 million inventor addresses. For the SCIE, the dataset comprises 9.1 million articles published during the same period containing 27.7 million listed author addresses.

The geocoding of addresses for this report is as follows. PCT inventor addresses were geocoded using the Environmental Systems Research Institute (ESRI) ArcGIS World Geocoder service.¹ When the ESRI address matches proved either insufficiently accurate or ambiguous, the city name in the address string was extracted and matched using records in the city level dataset from the GeoNames Gazetteer database.² This latter database gives the geolocation of cities around the globe and contains 48,000 geocoded cities. This same city matching approach was applied to all SCIE author addresses.

Overall, 96.4% of inventor addresses were geocoded at either the city level or a more accurate level, while 95.5% of scientific author addresses were geocoded at the city level. Annex Table 5 provides a summary of the geocoding results for the top 20 countries, which together account for the majority of inventor and scientific author addresses. As shown in the table, the coverage of geocoded addresses across all 20 countries is typically above 95%, only falling below 90% in one instance.

Addresses were clustered by applying the DBSCAN algorithm. This algorithm requires pre-defined radius and density parameters. As in previous years, a radius of 15 km and a density of 4,500 was applied. Equal weight was given to inventors and authors by expressing data points as a share of total inventor and author addresses, respectively. Given that the number of scientific articles far exceeds the number of patents, cluster identification based on the raw data points would have resulted in clusters shaped predominantly by the scientific author landscape.

The result was an initial list of 227 clusters. After review, neighboring clusters were merged if the edge of a cluster was within 3–5 km of another and where the co-author/co-inventor relationships were higher than they were for any other relationship with any other cluster or non-cluster points. A total of 22 clusters met these criteria, mergers reducing the overall number of clusters identified to 216.³

The remaining 216 clusters were then put into rank by counting the number of patents and scientific articles in a given cluster. Numbers were aggregated utilizing fractional counting, where counts reflect the share of a patent's inventors and an article's authors present in a particular cluster. In addition, mirroring the equal weighting approach described above, fractional counts are relative to the total numbers of patents and scientific articles.

To produce an intensity ranking, the European Commission's Global Human Settlement Layer (GHSL) population distribution data were matched geographically to the top 100 clusters identified in the overall ranking. Just as with inventor/author geocoded locations, this population data allowed us to define the total population of a cluster using a bottom-up approach. We chose to delimit a cluster's area as being all the space within 0.05 degrees of each inventor/author location. Overlaying the resultant cluster polygons on top of the population data and aggregating all points which lay within the polygon gave a total population estimate for each cluster.⁴ The clusters were then ranked by dividing the total S&T share by population.

