## 56 Estonia

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## 1 How have the numbers of researchers and technicians engaged in R&D per million people evolved relative to each other in Estonia between 1998 and 2016?

#### 1.1 Abstract

Using World Bank World Development Indicators (WDI) data, this study examines the evolution of Estonia's research and development (R&D) workforce between 1998 and 2016, focusing on the relationship between researchers and technicians per million people. These indicators provide insight into the structure, specialization, and capacity of Estonia's innovation system during a period of rapid technological and institutional transformation. Over the period, both researchers and technicians increased significantly, reflecting growing national investment in science, technology, and innovation. However, researchers grew at a faster rate than technicians, leading to a widening gap between the two groups. At the start of the period, researchers outnumbered technicians by roughly four to one; by 2016, that ratio had expanded to approximately six to one. This trend indicates that while both components of the R&D labor force expanded, Estonia's innovation system became increasingly research-intensive, emphasizing advanced scientific roles over technical support positions. These patterns highlight the country's transition toward a high-skill, knowledge-based economy, but also raise questions about balance and coordination within the R&D workforce.

#### 1.2 1. Question

How have the numbers of researchers and technicians engaged in R&D per million people evolved relative to each other in Estonia between 1998 and 2016?

- Researchers proxy: Researchers in R&D (per million people)
- Technicians proxy: Technicians in R&D (per million people)

#### 1.3 2. Data

- Source: World Bank World Development Indicators (WDI)
- Indicators:
  - Researchers in R&D (per million people)
  - Technicians in R&D (per million people)
- **Coverage**: Estonia, 1998–2016
- Notes: National-level data only

#### 1.4 3. Method

- 1. Filtered the dataset for Estonia and selected the two R&D workforce indicators.
- 2. Extracted relevant columns: Year, Indicator Name, and Value.
- 3. Pivoted the dataset to create a side-by-side chronological comparison of researchers and technicians.
- 4. Produced a dual-line time series plot to visualize growth patterns, relative magnitudes, and the widening gap between researchers and technicians.

(Analysis is descriptive; no causal inference applied.)

#### 1.5 4. Results

- Researchers in R&D (per million people): Increased by roughly 75% between 1998 and 2016, reflecting strong growth in higher education, research institutions, and innovation policy support.
- Technicians in R&D (per million people): Also increased, though at a slower pace—about 50% growth over the same period.
- Comparison: At the beginning of the period, researchers outnumbered technicians by approximately 4:1. By 2016, the ratio had expanded to about 6:1, showing that researcher growth outpaced technician growth, widening the gap despite overall increases in both categories.

(Figure 1. Estonia: Researchers vs. Technicians in R&D (per million people), 1998–2016)

(Table 1. Pivoted dataset summary)

## 1.6 5. Interpretation

- The overall upward trajectory of both researchers and technicians demonstrates Estonia's deepening commitment to innovation and scientific capacity-building during its post-independence transformation and integration into the European Union.
- The faster growth of researchers relative to technicians indicates a shift toward a more research-driven innovation model, emphasizing academic and scientific expertise.
- The widening gap suggests that technical support capacity did not expand at the same pace, potentially creating bottlenecks in applied research, experimentation, and technology transfer.
- This imbalance underscores the importance of maintaining complementarities between researchers and technicians to ensure that scientific discoveries translate efficiently into innovation and industrial application.

### 1.7 6. Limitations

- WDI data provide national-level aggregates and may not capture sectoral or institutional differences across universities, research institutes, and private firms.
- Cross-country comparability may be limited by differences in national definitions of "researchers" and "technicians."
- The descriptive analysis does not account for the impact of specific R&D policies, EU funding programs, or institutional reforms on workforce composition.

## 1.8 7. Next Steps / Extensions

- Investigate how changes in researcher-technician ratios relate to innovation outcomes, patent activity, and productivity growth.
- Compare Estonia's R&D labor structure with other Baltic and Nordic countries to contextualize regional innovation dynamics.
- Examine education and vocational training policies to assess whether technician training kept pace with research sector expansion.
- Explore more recent data to determine whether the researcher-technician gap continued to widen after 2016, or if policy adjustments have promoted greater balance within Estonia's R&D ecosystem.

```
[1]: # How have the numbers of researchers and technicians engaged in R&D peru
      →million people evolved relative to each other in Estonia between 1998 and
      →2016?
     import pandas as pd
     import matplotlib.pyplot as plt
     import os
     # Folders
     data_raw_folder = "data_raw/"
     data_clean_folder = "data_clean/"
     figures_folder = "figures/"
     # Load CSV
     filename = "science-and-technology_est_filtered.csv" # Filtered dataset with_
      ⇔only relevant rows
     df = pd.read_csv(os.path.join(data_raw_folder, filename))
     # Keep only needed columns
     df = df[["Year", "Indicator Name", "Value"]]
     # Convert Year and Value to numeric, drop invalid rows
     df["Year"] = pd.to_numeric(df["Year"], errors="coerce")
     df["Value"] = pd.to_numeric(df["Value"], errors="coerce")
     df = df.dropna(subset=["Year", "Value"])
     # Pivot indicators into separate columns
     df_pivot = df.pivot(index="Year", columns="Indicator Name", values="Value").
      →reset index()
     df_pivot = df_pivot.sort_values("Year")
     print("Pivoted Estonia dataset:")
     display(df_pivot)
     # Interpolate missing values for smooth plotting (optional)
     df_plot = df_pivot.interpolate(method='linear')
```

```
# Plot the indicators
plt.figure(figsize=(10,6))
plt.plot(df_plot["Year"], df_plot["Researchers in R&D (per million people)"],
        marker='o', linestyle='-', label="Researchers in R&D (per million_
 →people)")
plt.plot(df_plot["Year"], df_plot["Technicians in R&D (per million people)"],
        marker='o', linestyle='-', label="Technicians in R&D (per million_
 →people)")
plt.title("Estonia: Researchers vs Technicians in R&D (per million people)⊔
 plt.xlabel("Year")
plt.ylabel("Number")
plt.legend()
plt.grid(True)
plt.tight_layout()
plt.savefig(os.path.join(figures_folder,_

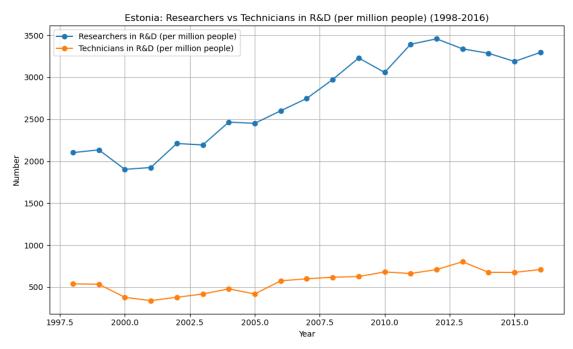
¬"estonia_researchers_vs_technicians_in_r&d.png"))
plt.show()
# Save cleaned CSV
df_pivot.to_csv(os.path.join(data_clean_folder,_

¬"estonia_researchers_vs_technicians_in_r&d"), index=False)
```

#### Pivoted Estonia dataset:

Indicator Name	Year	Researchers	in R	₹&D	(per million people) \
0	1998				2102.379883
1	1999				2135.264160
2	2000				1902.600586
3	2001				1924.963013
4	2002				2210.916748
5	2003				2193.733398
6	2004				2465.568848
7	2005				2451.187744
8	2006				2600.754150
9	2007				2747.582031
10	2008				2972.930664
11	2009				3229.675293
12	2010				3057.782959
13	2011				3392.805908
14	2012				3457.755615
15	2013				3338.395996
16	2014				3285.593506
17	2015				3188.491943
18	2016				3296.372559

Indicator Name	${\tt Technicians}$	in R&D	(per million people)
0			538.98428
1			534.26833
2			378.81170
3			339.14624
4			378.95991
5			417.36197
6			479.62510
7			418.25017
8			575.08617
9			598.82645
10			617.87679
11			625.68528
12			680.87930
13			662.04266
14			708.90951
15			802.08512
16			676.14947
17			675.11832
18			710.21109



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