

# Visualizing Innovation in Antibiotics

## Project Proposal for Data Visualization (EPPS 6356)

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### 1 Topic: Innovation in Antibiotics

I propose a project to visualize the innovation policy problem underlying Bacterial Antimicrobial Resistance (hitherto, AMR).

The AMR problem has been widely recognized as a global public health “threat” (Centers for Disease Control and Prevention (U.S.) 2019) with 1.27 million deaths worldwide in the year 2019 being attributed to it (Murray et al. 2022). These deaths are caused by the fact that bacteria have evolved to combat against currently available antibiotic drugs while innovation in antibiotics remains insufficient to invent new antibiotics to keep up with the pace of bacterial evolution (Spellberg et al. 2004; Freire-Moran et al. 2011; Zhang et al. 2016; Clerk 2015). Several policy incentives are currently being explored to ‘pull’ and/or ‘push’ antibiotic drug innovation further (McDonnell et al. 2023; Renwick, Brogan, and Mossialos 2016) with attempts being made to estimate the appropriate size of such incentives (Towse and Bonnifield 2022; Outtersson 2021). To aid in these attempts, I propose to make granular, reproducible visualizations using public available data sources at the level of individual molecular structures found in worldwide patents with potential/known antibiotic action.

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There are a couple of well-cited examples of existing dashboards that aim to provide evidence for antibiotic innovation. First is the dashboard by Global AMR R&D Hub (2025). Their visualizations do not document patents, and don't provide a visual comparison of innovation across different molecules (which is necessary for my research statement below). Second is the dashboard by Susan Hawes et al. (2025) which explicitly documents patents. But it uses the CPC system used by patent examiners instead of the molecular structures mentioned in the patents as I propose to do here. This can result in selection biases since firms can have strategic reasons not to reveal the therapeutic activity of a new molecule to the patent examiner given that many drugs can be patented before finishing the clinical trial process.

## **2 Research statement**

While this project is aimed to provide descriptive evidence on patented molecules with potential antibiotic action, it brings me one step closer to address a bigger research question: *How much worse are the product market outputs of antibiotic patents than comparable patents with similar inputs?* Notice that an answer to this question would clearly demonstrate the extent of the innovation insufficiency in antibiotics, and this can be useful to reckon the size of required incentives.

Methodologically, this research question faces two challenges. First is the attribution challenge of classifying patents attributable to the antibiotic class. Second is finding non-antibiotic patents similar to antibiotic patents (i.e) the challenge of identifying the right comparison group. Notice that both problems can be resolved by identifying the molecules in the patent. Once the molecules in the patent are known, I can use it to classify patents as belonging to the antibiotic class. Then, using well-defined measures of molecular similarity, I can find patents with molecules similar to antibiotic molecules.

Yet another challenge here is ensuring comparisons only among patents with 'similar inputs'. This is relegated for a future research project.

Figure 1: Visualization of AMR burden taken from Murray et al. (2022)

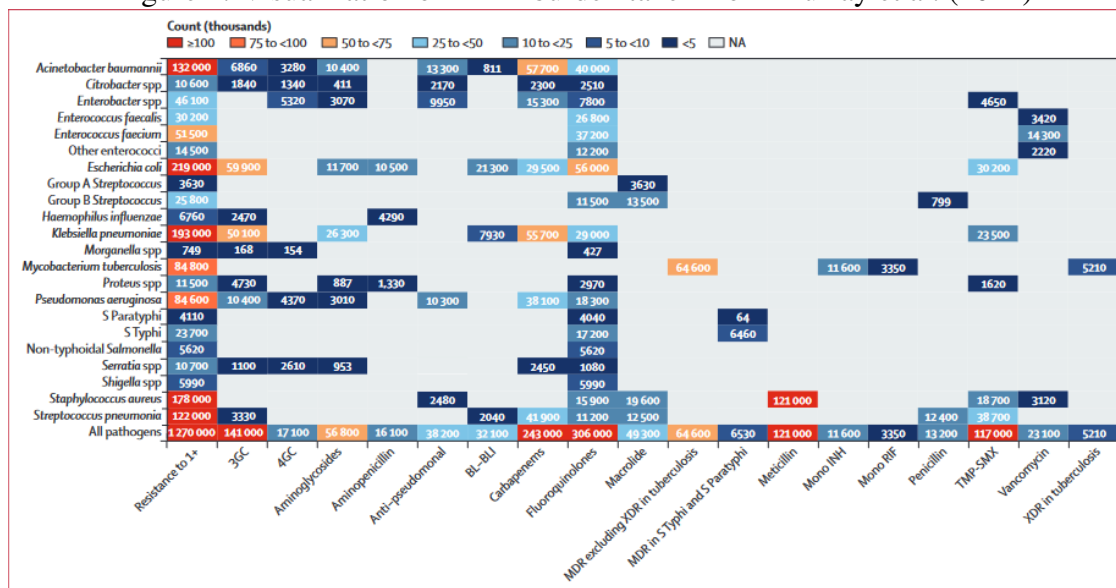


Figure 6: Global deaths (counts) attributable to bacterial antimicrobial resistance by pathogen-drug combination, 2019

For this figure, only deaths attributable to resistance, not deaths associated with resistance, are shown due to the very high levels of correlation for resistance patterns between some drugs. 3GC=third-generation cephalosporins. 4GC=fourth-generation cephalosporins. Anti-pseudomonal=anti-pseudomonal penicillin or beta-lactamase inhibitors. BL-BLI=β-lactam or β-lactamase inhibitors. MDR=multidrug resistance. Mono INH=isoniazid mono-resistance. Mono RIF=rifampicin mono-resistance. NA=not applicable. Resistance to 1+=resistance to one or more drug. S Paratyphi=Salmonella enterica serotype Paratyphi. S Typhi=Salmonella enterica serotype Typhi. TMP-SMX=trimethoprim-sulfamethoxazole. XDR=extensive drug resistance.

### 3 Method

The proposal here can be brought to fruition using two data sources both of which are publicly available to download. First is SureChEMBL (Papadatos et al. 2016), a cheminformatics database that uses machine learning techniques to annotate the molecules mentioned in different sections (abstract, description, claims, attachments) of worldwide patents. Secondly, I use AntibioticDB (Farrell et al. 2018) to identify molecules with potential antibiotic action. AntibioticDB also has self-reported data on which stage of the clinical trial pipeline each molecule was last observed. Both SureChEMBL and AntibioticDB use the International Chemical Identifier (InChi) key to identify individual molecules, helping me join them both.

I propose to use these two datasets to make a visualization comparable to the visualization of AMR burdens provided in Murray et al. (2022). This visualization is shown in Figure 1. Specifically, I hope to make a matrix visualization that maintains the same x axis as Figure 1. But instead of the numbers/colors describing the mortality, I want to describe innovation proxied by patents. I want my figure to look as similar as possible to Figure 1 so that I can unambiguously motivate why it is significant for policymakers to understand the innovation problem in a granular way. That is, I

want policymakers to be able to see if innovation is worsen in those drug molecules that are known to cause higher AMR mortality.

An obstacle that I won't be able to overcome in this project is plotting the data using the same y-axis as Figure 1. This is because SureChEMBL doesn't annotate pathogens mentioned in patents. But even if such data were available, innovation in antibiotics can't be presented with the same grammar. Given that many antibiotic molecules can be broad-spectrum (i.e) can have action against multiple species of bacteria it will be difficult to 'fractionalize' innovation across all of them. This will require me to identify other policy-relevant y-axis.

In addition to the matrix visualization I have described, I hope to make visualizations describing the patent assignees and inventors (i.e) the firms that own the molecules and the researchers who invented them. Given the global coverage of patents in SureChEMBL it should also be possible to make choropleths describing the spatial variation in patents across countries. If time permits, I can also use data on patents specifically in the US, a major hub of pharmaceutical innovation, sourced from the US Patents and Trademarks Office (USPTO) to understand spatial location of inventors across US states.

Given that this project is aimed at enabling policymakers to make better estimates of the size of policy incentives, reproducibility of analyses will be a cornerstone of this project. With this in mind, throughout this project, I propose to use the ggplot2 R package (Wickham 2016), which uses the Grammar of Graphics framework of Leland Wilkinson (2005). I propose to deploy my visualization using the Shiny framework (Chang et al. 2025). Notice that all the tools here are open-source. In addition, I hope to have my code in a GitHub repository. All this, in conjunction with the public nature of the two datasets, should make this project entirely reproducible by external stakeholders.

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