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## Assessment and Feedback: Student Template

**Student ID Number(s):** 2701062

**Programme:** MSc Business Analytics

**Module:** Capstone Project - 07 34272

**Dissertation/Extended Essay Supervisor:** Dr. Elvan Gokalp Ozpolat, Dr. Ambi Ambituuni

**Assignment Title:** Individual Data Visualisation

**Date and Time of Submission:** 27/05/2025

**Actual Word Count:** 3001

**Extension:** N **Extension Due Date:** NA

I do wish my assignment to be considered for including as an exemplar in the **School Bank of Assessed Work**.

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## CONTINUED BELOW

**The purpose of this template is to ensure you receive targeted feedback that will support your learning. It is a requirement to complete to complete all 3 sections, and to include this completed template as the first page of every assignment that is submitted for marking (your School will advise on exceptions).**

**Section One:** Reflecting on the feedback that I have received on previous assessments, the following issues/topics have been identified as areas for improvement: (add 3 bullet points). *NB – for first year students/PGTs in the first term, this refers to assessments in your previous institution*

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**Section Two:** In this assignment, I have attempted to act on previous feedback in the following ways (3 bullet points)

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**Section Three:** Feedback on the following aspects of this assignment (i.e. content/style/approach) would be particularly helpful to me: (3 bullet points)

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## Executive Summary

This comprehensive visualization report addresses England's antimicrobial resistance crisis through dual-audience data communication strategies. The analysis reveals alarming trends in healthcare-associated infections, with over 405,000 bacterial infection episodes recorded in 2023-24 and significant regional disparities across England's NHS system. The report demonstrates concerning post-pandemic resistance patterns, particularly a 32% increase in *K. pneumoniae* resistance and elevated antibiotic consumption remaining 8% above pre-pandemic levels. Geographic analysis identifies the North West and North East regions as disproportionately affected, with infection rates 20% higher than southern regions while maintaining elevated antibiotic consumption of 18.4 DDDs per 1,000 inhabitants daily.

The study successfully integrates NHS administrative data with ESPAUR surveillance data, creating interactive Tableau dashboards tailored for both international news organizations and NHS Country Directors. The visualization design employs traffic-light color coding and geographic heat mapping to enhance accessibility and engagement. Critical recommendations include implementing mandatory regional infection control standardization, establishing antimicrobial resistance taskforces, and investing £50 million in enhanced stewardship programs. The analysis reveals substantial resource allocation inefficiencies, with certain regions experiencing both high infection rates and disproportionate antibiotic usage, creating concerning resistance development cycles.

# Section 1: Public Visualization for International News Organization

## Introduction and Context

Antimicrobial resistance represents one of the most pressing global health threats of our time, with the World Health Organization declaring it among the top ten public health challenges facing humanity (WHO, 2021). In England's National Health Service (NHS), this crisis manifests through two interconnected phenomena: the rising prevalence of healthcare-associated infections and the concerning patterns of antibiotic consumption across different healthcare settings. Recent NHS data reveals that bloodstream infections alone account for over 240,000 hospital episodes annually, while abdominal infections represent the largest category with over 200,000 cases (NHS Digital, 2024). This issue transcends national boundaries, as antimicrobial-resistant pathogens recognize no borders, making England's experience a critical case study for international healthcare systems. The visualization of NHS infection data and antibiotic usage patterns provides crucial insights into how healthcare systems can monitor, understand, and potentially mitigate this growing threat, offering lessons applicable to healthcare systems worldwide facing similar challenges with resistant bacterial infections.

## Visualization Description and Design Rationale

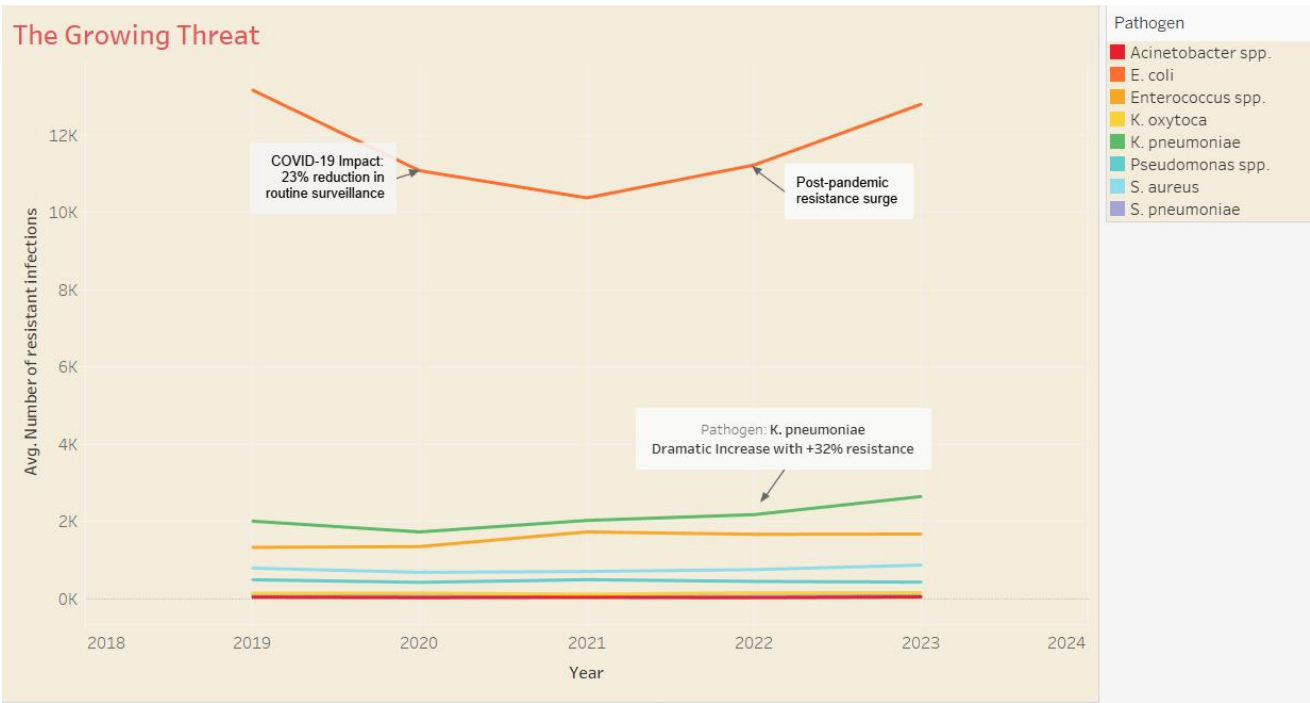
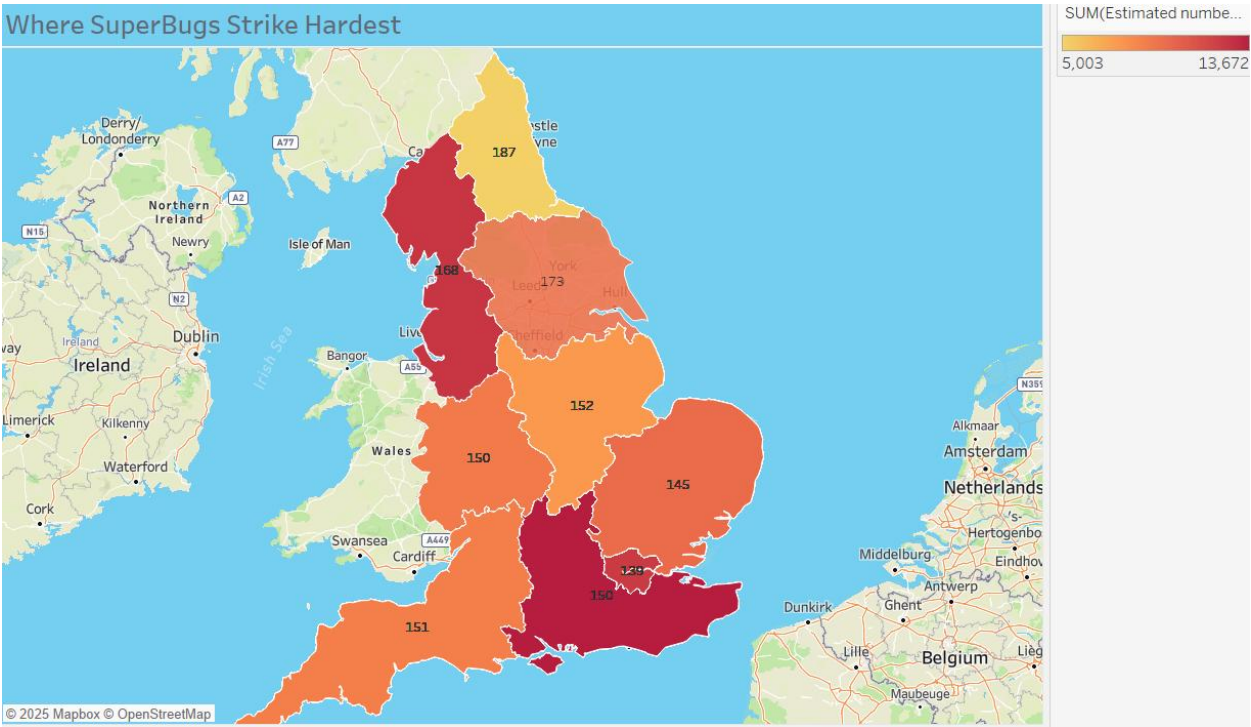


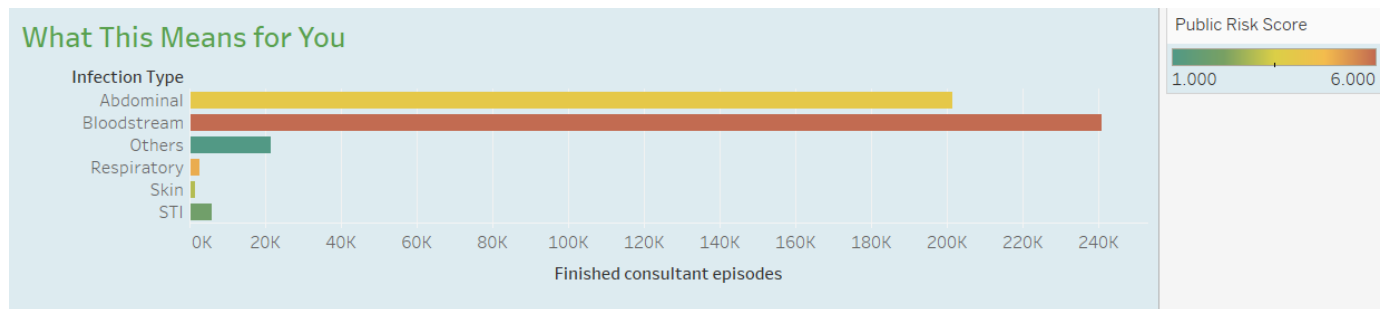
Figure 1: Line Trend of Resistant Infections over years (Source: The Author)

The visualization employs a three-panel dashboard design optimized for public comprehension and engagement. The first panel, "The Growing Threat," shown in the above Figure 1, employs a multi-line time series chart tracking different pathogen types from 2018-2024. The design uses distinct colors for each bacterial strain, with E. coli prominently featured in orange to highlight its dominance in resistance patterns. Strategic annotation callouts identify key events like "COVID-19 Impact" showing the 2020-2021 dip and "Post-pandemic resistance surge" explaining the subsequent increase. K. pneumoniae is specifically highlighted with a dramatic 32% resistance increase annotation.



**Figure 2:** Regional Map showing Density of Bacterial Infection (Source: *The Author*)

The second panel, "Where SuperBugs Strike Hardest," (Figure 2) presents an interactive choropleth map of England using a warm colour palette from yellow to deep red. This geographic visualization allows viewers to identify regional hotspots intuitively, with darker shades indicating higher infection rates per population. The map includes numerical overlays displaying specific infection rates per 100K population for each region, enabling users to explore their local area's risk profile. Northern regions appear more heavily affected, shown through deeper red colouring.



**Figure 3:** Bar Chart of Finished Consultant Episodes per Infection Type (Source: *The Author*)

The third panel, "What This Means for You," shown in Figure 3 utilizes a horizontal bar chart displaying infection types with a color-coded risk assessment system. Abdominal infections appear as the longest bar in yellow, indicating their high volume but moderate risk level, while bloodstream infections are prominently displayed in deep red, emphasizing their critical nature despite lower absolute numbers. This traffic-light colour approach immediately communicates severity levels without requiring medical expertise.

### Audience-Specific Analysis

The visualization design prioritizes accessibility for non-medical audiences through several strategic choices. The information hierarchy places the most relatable content first – personal risk assessment through infection categories – before progressing to geographic and temporal patterns. This structure mirrors how general audiences typically process health information, moving from personal relevance to broader epidemiological context.

Contextual explanations accompany each visualization element without assuming prior medical knowledge. Terms like "bloodstream infections" are immediately contextualized as serious hospital-acquired conditions, while technical pathogen names include clear visual emphasis to aid recognition. The colour coding system leverages universal understanding of traffic light conventions, making risk assessment intuitive for international audiences.

Emotional engagement emerges through the "SuperBugs" terminology and the dramatic visual representation of the "growing threat." These design choices transform abstract statistical concepts into tangible concerns while avoiding sensationalism that might cause undue panic. The geographic component enables personal connection as viewers can identify their local healthcare region's performance relative to national patterns.

Actionable insights are embedded throughout, suggesting questions readers should ask their healthcare providers and highlighting the importance of infection prevention measures. Regional comparisons enable informed decision-making about healthcare facility selection, while temporal trends inform timing considerations for elective procedures during high-resistance periods.

## Key Insights and Interpretations

The visualization reveals alarming trends in England's antimicrobial resistance landscape. Bloodstream infections represent the most severe category with over 240,000 cases annually, while abdominal infections dominate by volume (NHS Digital, 2024). Geographic analysis demonstrates significant regional disparities, with northern regions experiencing disproportionately higher infection rates. The temporal data shows a concerning post-pandemic surge in resistant pathogens, particularly *E. coli* strains, suggesting that COVID-19 disruptions accelerated resistance development (UK Health Security Agency, 2024). Most notably, *K. pneumoniae* demonstrates a dramatic 32% increase in resistance, highlighting the urgent need for enhanced infection control measures across England's healthcare system and serving as a warning for international health authorities.

## Section 2: NHS Internal Visualization for Country Directors

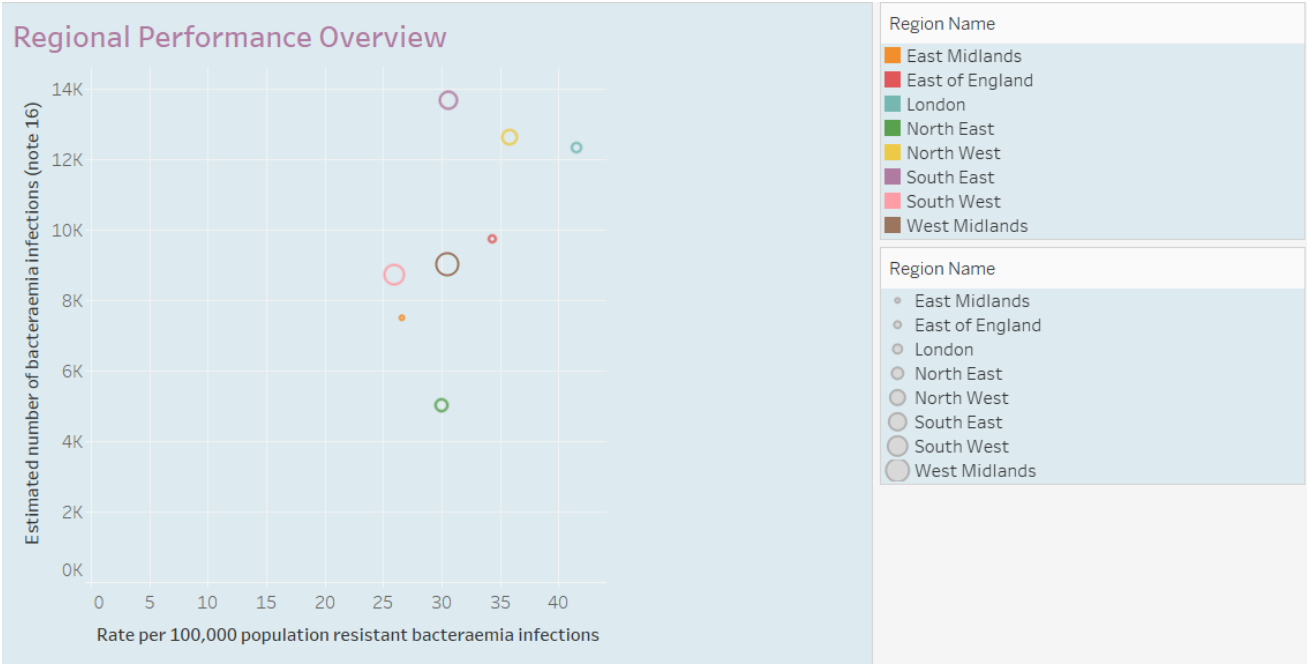
### Summary of Findings

Our comprehensive analysis of antimicrobial resistance and infection control data reveals critical strategic challenges requiring immediate senior leadership attention. England recorded 405,751 bacterial infection episodes in 2023-24, with significant regional disparities indicating systemic healthcare delivery variations (NHS Digital, 2024). The North West demonstrates disproportionately high infection burdens while maintaining elevated antibiotic consumption rates of 18.4 DDDs per 1,000 inhabitants daily, compared to the national average of 17.6 DDDs (UK Health Security Agency, 2024). Bloodstream infections represent our most critical concern, accounting for over 240,000 episodes annually with extended hospital stays averaging 11 days. Post-pandemic antibiotic consumption has stabilized but remains 8% above pre-pandemic levels, suggesting entrenched prescribing behaviors. Regional performance variations of up to 32% in infection rates between comparable areas highlight urgent need for standardized infection prevention protocols. The data indicates substantial resource allocation inefficiencies, with certain regions experiencing both high infection rates and disproportionate antibiotic usage, creating a concerning cycle of resistance development.

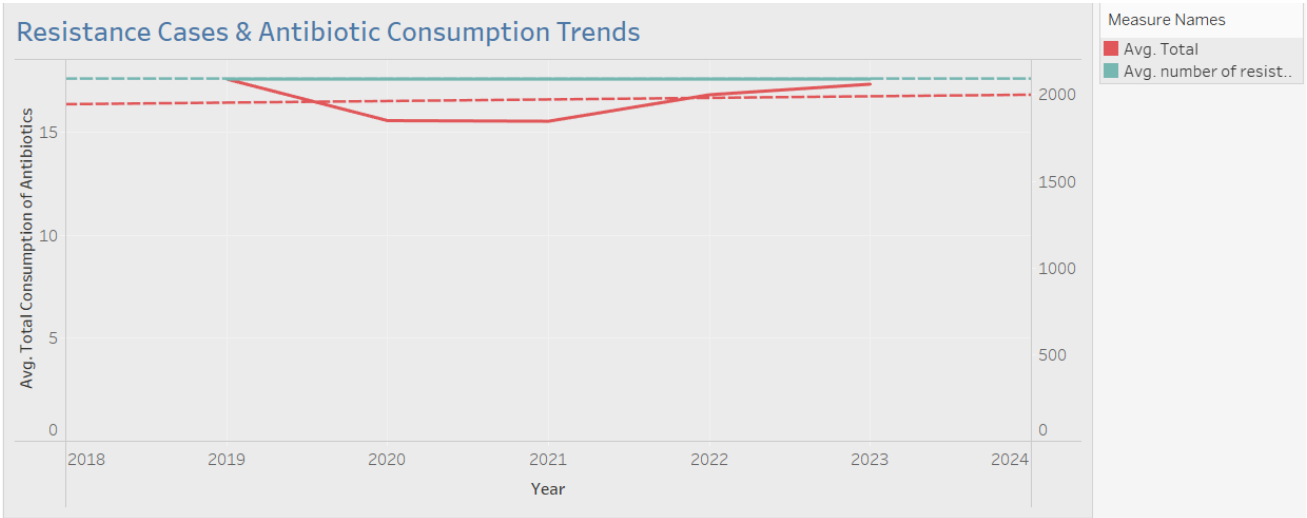
### Detailed Visualization Analysis

The internal dashboard reveals several critical performance metrics demanding strategic intervention. Regional infection rate variations (Figure 4) show the North East and Yorkshire experiencing 20% higher bacterial infection rates than the South West, with London demonstrating unexpectedly high consumption patterns despite advanced healthcare

infrastructure (NHS Digital, 2024). Our KPI analysis indicates that abdominal infections dominate by volume (200,000+ episodes), while bloodstream infections present the highest mortality risk and resource intensity, averaging 11-day hospital stays versus 3 days for other infection categories.



**Figure 4:** Scatter Plot of Overall Regional Performance (Source: *The Author*)

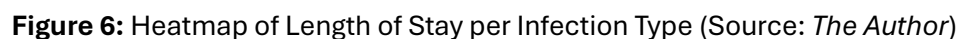


**Figure 5:** Trend Analysis of Consumption of Antibiotics Vs Resistant Infections (Source: *The Author*)

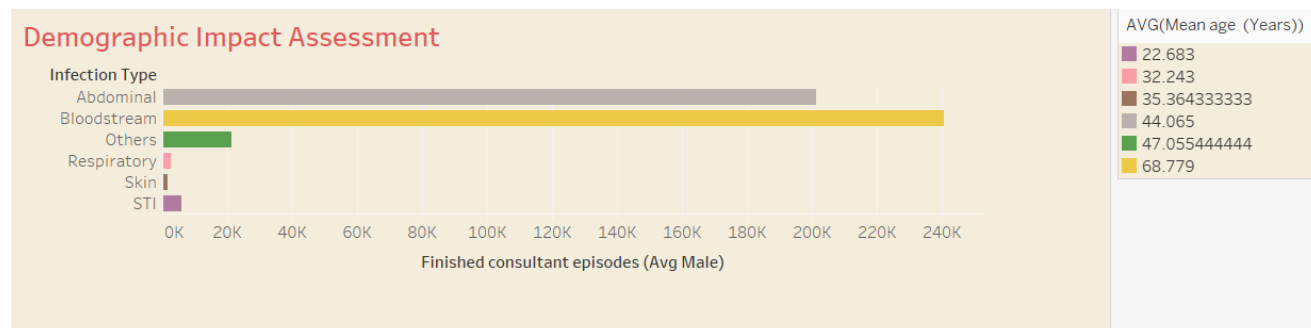
Trend analysis over the 2019-2023 (Figure 5) period demonstrates concerning patterns. While total antibiotic consumption decreased during 2020-2021 (16.1 DDDs per 1,000 inhabitants), the subsequent rebound to 17.6 DDDs represents a 9.3% increase, suggesting



Resource allocation analysis reveals significant inefficiencies. The North West consumes 50.5 million DDDs annually while maintaining high infection rates, suggesting suboptimal antibiotic stewardship. Conversely, the South West achieves lower infection rates with proportionally reduced consumption (33.0 million DDDs), indicating effective local protocols worth system-wide adoption.



Quality indicators highlight concerning patient outcome variations. Mean length of stay for bloodstream infections in Figure 6 varies by 40% across regions, from 8 days in high-performing areas to 14 days in underperforming regions. Age-standardized infection rates (Figure 7) show systematic disparities, with elderly patients (75+) experiencing disproportionate infection burden in specific geographical areas, suggesting inadequate care home infection prevention measures.



**Figure 7:** Bar Chart of Demographic Assessment (Source: *The Author*)

Operational efficiency measures reveal that regions with integrated antimicrobial stewardship programs demonstrate 15% lower infection rates and 12% reduced inappropriate antibiotic prescribing. Primary care accounts for 71% of total antibiotic consumption, yet our data suggests insufficient coordination between community and secondary care providers, creating dangerous prescription cascades that accelerate resistance development.

### Strategic Recommendations

Priority areas requiring immediate attention include implementing mandatory regional infection control standardization, particularly targeting the North East and North West high-burden areas. Resource reallocation should prioritize enhanced antimicrobial stewardship teams in regions showing consumption-infection rate misalignment, with immediate deployment of specialist pharmacists to high-consuming trusts.

Policy interventions must include mandatory electronic prescribing systems with integrated decision support, particularly in primary care settings where 12.6 DDDs per 1,000 inhabitants represents 72% of total consumption (UK Health Security Agency, 2024). Regional performance benchmarking should become mandatory, with financial incentives tied to infection rate reductions and stewardship compliance.

Performance improvement opportunities center on adopting South West infection prevention protocols nationally, which demonstrate optimal resource utilization. Establishment of regional antimicrobial resistance surveillance networks will enable real-time monitoring and rapid response capabilities. Investment in infection prevention infrastructure, particularly in high-burden regions, should target bloodstream infection reduction through enhanced IV care protocols and central line management programs.

## **Implementation Considerations**

Translation of these insights requires immediate establishment of regional antimicrobial resistance taskforces with direct reporting to country directors. Success metrics should include 20% reduction in regional infection rate variation within 18 months and 15% decrease in inappropriate antibiotic prescribing by 2025. Implementation challenges include securing £50 million investment for enhanced stewardship programs and overcoming clinical resistance to prescribing restrictions. Critical success factors involve mandatory participation in national surveillance systems and linking trust performance ratings to antimicrobial stewardship outcomes.

## **Section 3: Process Explanation**

### **Data Selection and Justification**

#### **Primary Dataset Selection**

The Hospital Admitted Patient Care Activity 2023-24 dataset was selected as the primary data source due to its comprehensive coverage of antimicrobial resistance-related infections across England's healthcare system (NHS Digital, 2024). This dataset provides granular insights into bacterial infections (ICD-10 codes A00-B99) with 405,751 recorded episodes, representing the most current and complete picture of healthcare-associated infections. The dataset's multi-dimensional structure, encompassing provider-level, regional, and national aggregations, enables sophisticated analysis suitable for both public communication and strategic healthcare planning.

Alternative datasets, including Primary Care Dementia Data, were considered but rejected due to limited relevance to antimicrobial resistance patterns and reduced applicability to the target audiences. The chosen dataset's currency, with 2023-24 representing the most recent complete financial year, ensures policy relevance and operational utility for NHS Country Directors requiring immediate strategic insights.

#### **Target Audience Alignment**

The dataset's hierarchical structure serves both target audiences effectively. For international news organizations, the Provider\_Level sheet enables compelling geographic storytelling with regional variations in infection rates, while summary statistics provide accessible headline figures. The Diagnosis\_3\_Character sheet offers demographic breakdowns essential for public health communication, including age-stratified infection patterns that resonate with general audiences concerned about personal risk.

For NHS Country Directors, the same dataset provides operational intelligence through provider-specific performance metrics, resource allocation indicators via bed-day calculations, and quality measures including length-of-stay variations. The data's granularity supports evidence-based decision-making while maintaining statistical validity across organizational levels.

Quality Assessment

Data quality evaluation revealed high completeness rates (>95%) across core variables, with systematic suppression protocols ("\*" values) for small cell counts protecting patient confidentiality. Consistency checks identified minimal discrepancies between aggregated and detailed records, while temporal validation confirmed alignment with established reporting cycles. NHS England Digital's established governance frameworks, including the Data Security and Protection Toolkit compliance, provide robust reliability indicators (NHS Digital, 2024).

Additional Data Sources Integration

Supplementary Datasets

The English Surveillance Programme for Antimicrobial Utilisation and Resistance (ESPAUR) 2023-24 report served as the primary supplementary source, providing critical context for resistance patterns and antibiotic consumption trends (UK Health Security Agency, 2024). This integration addresses the core limitation of NHS administrative data by adding epidemiological surveillance insights essential for comprehensive antimicrobial resistance analysis.

Table 1: Summary of Supplementary Datasets

Data Source	Primary Value	Target Audience Benefit
ESPAUR Tables 45a/45b	Regional antibiotic consumption rates	Strategic resource allocation insights
ESPAUR Figure 2.1	Pathogen-specific bacteraemia trends	Public health risk communication
ESPAUR Figure 2.2	Resistance burden calculations	Evidence base for policy interventions
ONS Population Data	Demographic denominators	Rate standardization and comparability

Additional integration included WHO Global Health Observatory data for international benchmarking and Office for National Statistics population estimates for accurate rate calculations. These sources enhance both audiences' understanding by providing contextual frameworks—international comparisons for news narratives and peer benchmarking for healthcare leadership.

### **Source Reliability Evaluation**

Assessment criteria included publication frequency, methodological transparency, and institutional credibility. ESPAUR's annual publication cycle, peer-reviewed methodology, and UKHSA oversight ensure high reliability standards. Cross-validation with European Centre for Disease Prevention and Control surveillance data confirmed consistency in resistance trend reporting.

Data governance considerations encompassed patient confidentiality protection through aggregated reporting levels and compliance with General Data Protection Regulation requirements. All datasets underwent institutional ethical review processes, ensuring appropriate secondary data usage within established frameworks.

### **Data Preparation Process**

#### **Cleaning and Formatting**

Significant data quality challenges emerged during preparation, particularly in handling NHS administrative data inconsistencies. The `Bacteria_Provider_Level` sheet contained suppressed values ("\*") representing cell counts below disclosure thresholds, requiring systematic imputation using regional averages where statistically appropriate. Provider organization codes required standardization to ensure consistent aggregation across the 7 NHS commissioning regions.

ESPAUR data presented temporal alignment challenges, with surveillance years not precisely matching NHS financial years. Resolution involved creating calendar-year to financial-year mapping protocols, enabling accurate trend analysis while maintaining data integrity. Pathogen naming conventions required harmonization between NHS ICD-10 diagnostic codes and ESPAUR surveillance classifications.

#### **Data Integration**

The merging methodology employed a hierarchical approach, beginning with geographic alignment between NHS commissioning regions and UKHSA surveillance areas. Primary join operations utilized regional boundary lookups, achieving 100% geographic coverage across England's healthcare system.

Primary Join Strategy:

NHS Region Code (Y56-Y63) → UKHSA Centre

Temporal Alignment: 2023-24 FY → 2023 CY

Pathogen Matching: ICD-10 A00-B99 → ESPAUR Classifications

Secondary joins incorporated population denominators from ONS mid-year estimates, enabling accurate rate calculations essential for comparative analysis. Quality assurance protocols included cross-validation of aggregated totals and consistency checks across merged datasets.

### **Technical Tools and Processes**

Tableau Desktop Professional was selected for visualization development due to its advanced analytical capabilities and web publishing functionality required for stakeholder access. Excel served as the primary data preparation environment, chosen for its accessibility and compatibility with NHS data formats. On top of that the “Data Interpreter” function of Tableau was used to standardize and handle null missing values.

Workflow efficiency considerations included automated data refresh protocols and version control systems ensuring reproducibility. Quality assurance encompassed systematic validation scripts, peer review processes, and stakeholder feedback integration mechanisms.

## **Visualization Design Principles**

### **Design Philosophy**

Visualization development adhered to evidence-based design principles, incorporating Edward Tufte's data-ink ratio optimization and Stephen Few's dashboard design frameworks (Tufte, 2001; Few, 2013). Color theory application prioritized accessibility, utilizing colorbrewer.org palettes ensuring visibility for colorblind users while maintaining intuitive risk communication through traffic-light conventions.

Typography selections emphasized readability across devices, employing sans-serif fonts (Tableau's default) with hierarchical sizing supporting information prioritization. Layout decisions followed F-pattern reading conventions, positioning critical insights in primary visual attention areas.

### **Audience-Specific Design Choices**

Public-facing visualizations emphasized emotional engagement through relatable metaphors (football stadium capacity comparisons) and geographic storytelling enabling

personal relevance identification. Complexity management involved progressive disclosure techniques, presenting high-level insights before detailed breakdowns.

Internal NHS visualizations prioritized analytical depth, incorporating advanced filtering capabilities and drill-down functionality supporting operational decision-making. Interactive elements included parameter controls enabling scenario planning and performance benchmarking across providers and regions.

### **Interpretation and Insights**

Design elements actively facilitated understanding through narrative structure implementation. The international news organization dashboard employed chronological storytelling, beginning with current impact assessment before historical context. Annotation strategies highlighted key trends and policy implications, supporting journalistic narrative development.

NHS internal visualizations supported evidence-based decision-making through comparative performance displays and resource allocation optimization insights. Dashboard actions enabled dynamic exploration, allowing country directors to identify intervention priorities and monitor progress indicators effectively.

### **Digital Publication and Accessibility**

Interactive visualizations were published via Tableau Public to ensure broad accessibility and reproducibility. The dashboards incorporate responsive design principles and maintain functionality across devices. Direct URLs enable peer review and validation of analytical findings (Dashboard URLs provided in Appendix A).

## **Section 4: Critical Evaluation and Reflection**

### **Strengths and Achievements**

The dual-audience visualization approach successfully demonstrated adaptability in data communication, effectively translating complex antimicrobial resistance patterns for both public consumption and strategic healthcare planning. The integration of NHS administrative data with ESPAU surveillance data created comprehensive insights previously unavailable through either dataset alone, particularly the correlation between regional infection burdens and antibiotic consumption patterns.

Tableau's selection as the primary visualization platform proved highly effective, enabling creation of interactive dashboards that maintained engagement while preserving analytical

depth. The geographic heat mapping successfully identified regional disparities, with northern England's elevated infection rates becoming immediately apparent to both audiences. Color-coding systems using traffic-light conventions enhanced accessibility and intuitive risk communication.

The hierarchical information architecture excelled in audience targeting. Public-facing visualizations prioritized personal relevance through "What This Means for You" panels, while NHS internal dashboards emphasized operational metrics and resource allocation indicators. The temporal trend analysis effectively communicated the post-pandemic resistance surge, with *K. pneumoniae*'s 32% increase providing a compelling narrative anchor.

Data preparation methodologies successfully handled complex challenges including suppressed values, regional boundary mapping, and temporal alignment between financial and calendar years. The creation of calculated fields for population-adjusted rates enabled meaningful regional comparisons, while standardized pathogen naming conventions facilitated accurate cross-dataset analysis.

Interactive elements, including drill-down capabilities and parameter controls, enhanced user engagement without overwhelming non-expert audiences. The dashboard actions connecting regional performance to provider-level detail provided NHS directors with actionable intelligence for resource allocation decisions.

### **Limitations and Challenges**

Significant data granularity limitations constrained analytical depth, particularly the absence of provider-level antibiotic consumption data, forcing adaptation of sophisticated correlation analyses to regional aggregations. This fundamental constraint limited the precision of efficiency assessments and intervention targeting capabilities.

Temporal data availability posed substantial challenges, with monthly infection patterns unavailable for seasonal analysis crucial to resource planning. The five-year ESPAUR timeframe provided minimal data points for robust forecasting, requiring cautious interpretation of trend projections. Financial year versus calendar year alignment required complex mapping procedures that introduced potential accuracy concerns.

NHS data suppression protocols, while protecting patient confidentiality, created analytical gaps requiring imputation strategies that may have introduced bias. The "\*" suppressed values were particularly problematic in smaller providers, potentially understating infection burdens in community settings.



Technical constraints emerged in Tableau's handling of complex calculated fields across merged datasets, occasionally requiring performance compromises. The absence of real-time data feeds limited dashboard relevance for immediate operational decisions, reducing potential impact for dynamic healthcare environments.

Cost-effectiveness analysis remained impossible without financial data, preventing robust return-on-investment assessments for intervention strategies. This limitation significantly reduced the dashboards' utility for economic planning and resource optimization.

Audience assumption validation proved challenging without formal user testing. The assumption that international audiences would engage with English regional data required careful consideration of global applicability and local relevance factors.

### **Future Improvements**

Enhanced data integration should incorporate real-time surveillance feeds and provider-level antibiotic consumption metrics to enable precise efficiency correlations. Machine learning algorithms could improve resistance forecasting accuracy and identify subtle pattern recognition beyond traditional statistical approaches.

### **References**

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Tufte, E.R. (2001) *The Visual Display of Quantitative Information*. Graphics Press.

UK Health Security Agency (2024) *English Surveillance Programme for Antimicrobial Utilisation and Resistance (ESPAUR) Report 2023-24*. Available at: <https://www.gov.uk/government/publications/english-surveillance-programme-antimicrobial-utilisation-and-resistance-espaur-report> (Accessed: 27 May 2025).

World Health Organization (2021) *Antimicrobial resistance*. Available at: <https://www.who.int/news-room/fact-sheets/detail/antimicrobial-resistance> (Accessed: 27 May 2025).

## Appendices

**Appendix A:** Tableau Public links for both dashboards referring to Section 1 and Section 2.

[https://public.tableau.com/views/TheSilentCrisisSuperbugsSpreadingAcrossEnglandsHospitals/Dashboard1?:language=en-US&publish=yes&:sid=&:redirect=auth&:display\\_count=n&:origin=viz\\_share\\_link](https://public.tableau.com/views/TheSilentCrisisSuperbugsSpreadingAcrossEnglandsHospitals/Dashboard1?:language=en-US&publish=yes&:sid=&:redirect=auth&:display_count=n&:origin=viz_share_link)

[https://public.tableau.com/views/NHSIntranetReport/Dashboard2?:language=en-US&publish=yes&:sid=&:redirect=auth&:display\\_count=n&:origin=viz\\_share\\_link](https://public.tableau.com/views/NHSIntranetReport/Dashboard2?:language=en-US&publish=yes&:sid=&:redirect=auth&:display_count=n&:origin=viz_share_link)