***Achromobacter xylosoxidans* isolates exhibit genome diversity, variable virulence, high**
levels of antibiotic resistance and potential intrahost evolution.

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Abstract

Achromobacter xylosoxidans is an emerging pathogen characterized by high levels of
antibiotic resistance and increasing infection rates worldwide. This motile, opportunistic
pathogen is widely distributed in the environment and can cause various infections, including
pneumonia, bacteremia, endocarditis, meningitis, and others. In this study, we analyzed the
population structure, antibiotic resistance profiles, and virulence factors of over 200 publicly
available genomes. Core genome analysis revealed that *A. xylosoxidans* is highly adaptable,
possessing a relatively small core genome. Antibiotic susceptibility testing of isolates from the
United States revealed high resistance to multiple antibiotics including
ceftolozane/tazobactam, aztreonam, cefepime, and ciprofloxacin. Our data show that
imipenem/relebactam (IMR) is not more effective against *A. xylosoxidans* than imipenem (IMI)
alone, indicating that relebactam does not inhibit β-lactamase activity in Achromobacter. The
species features multiple secretion systems, including the Type III secretion system (T3SS) of
the YscN family, which is similar to those found in *Bordetella pertussis* and *Pseudomonas*
aeruginosa. Isolates collected from the same patients showed dynamic changes in
cytotoxicity, flagella motility, biofilm and antibiotic resistance suggesting its dynamic
adaptation to host environment. Intra-host evolved isolates, NIH-010 and NIH-016,
demonstrated the loss of flagella motility and variable cytotoxicity while exhibiting increased
antibiotic resistance and enhanced biofilm formation. Sequence analysis suggests that NIH-
016-3 has tyrosine to histidine mutation at position 330 (Y330H) near the FlhF Guanosine
triphosphate (GTP)-binding domain that may affect flagellar assembly. Interestingly, virulence
assays showed significant variation in the ability of different *A. xylosoxidans* isolates to induce

cell death in *in vitro* models, suggesting its dynamic adaptation to host environment. These
findings highlight the complexity of this pathogen group and underscore the urgent need for
further research into its mechanisms of antibiotic resistance and virulence.

Impact statement

This study provides a comprehensive examination of *Achromobacter xylosoxidans*, an
emerging pathogen of global concern due to its high antibiotic resistance and increasing
clinical relevance. By analyzing over 200 genomes, we offer critical insights into the population
structure, resistance mechanisms, and virulence factors of this species. The identification of a
relatively small core genome underscores the adaptability of *A. xylosoxidans* and its potential
for genomic plasticity. The existence of multiple secretion systems (T1SS to T6SS) highlights
the great capacity of *A. xylosoxidans* as a pathogen. Variations in virulence among *A.
xylosoxidans* isolates indicate the complexity of this pathogen and underscore the need for
further studies on its virulence mechanisms. The dynamic within-host evolution including the
loss of motility-associated systems and the acquisition of enhanced antibiotic resistance and
biofilm formation, adds new dimensions to our understanding of its pathogenesis. Flagella, an
important virulence mechanism, exhibit variable motility in different isolates, suggesting the
bacteria use flagella as an adaptive mechanism. This work also showed that *A. xylosoxidans*
are resistant to relebactam, which are effective in other bacteria. These findings emphasize
the urgent need for targeted therapeutic strategies to combat this opportunistic pathogen.

1.   Introduction

The genus *Achromobacter* is a non-lactose fermenting, Gram-negative, motile bacterium
recognized as an emerging pathogen in cystic fibrosis (CF) patients (1). It belongs to the β-
proteobacteria class. While *Achromobacter* species are ubiquitously found in water and soil,
and can grow in the plant rhizosphere, *A. xylosoxidans* isolates are primarily clinically derived
(2). This bacterium was first isolated in 1971 from an ear infection (3). *A. xylosoxidans* can
cause a wide range of infections, including pneumonia, bacteremia, endocarditis, meningitis,
and others (4, 5). These infections, which predominantly affect immunocompromised
individuals often lead to severe outcomes (4). Colonization with *A. xylosoxidans* is considered
a marker of disease severity in CF patients (6) and is associated with a decline in respiratory
function (7) and with a greater risk of death in lung transplant recipients (8). The persistence
of *A. xylosoxidans* in both extreme environments and host organisms may be attributed to its
genetic diversity and adaptability, although this aspect remains poorly understood (9). Earlier
studies examining the genomic diversity of this bacterial genus across six genomes revealed a

higher percentage of accessory genes in *Achromobacter spp*. (10). Recent virulence studies
have utilized various clinical isolates to explore its pathogenic mechanisms (11-13). As of the
time of this study, nearly 200 genomes of *A. xylosoxidans* were available on NCBI. However,
comprehensive analyses of the differences among isolates with respect to their genomic
content, virulence potential, and antibiotic resistance profiles are lacking.

Clinical isolates possess unique genomic features that differentiate them from environmental
isolates and contribute to virulence, such as the type III secretion system (T3SS), a
polysaccharide island, adhesion-related proteins, and several putative toxins (14). In addition
to these genetic elements, the Type VI Secretion System (T6SS) has been recently identified
to contribute to virulence alongside the T3SS (15, 16). The T3SS enables bacteria to deliver
effector proteins directly into the cytosol of animal or plant host cells. These systems have
diversified into seven families, spreading across Gram-negative bacteria through horizontal
gene transfer (17). The T6SS facilitates bacterial competition and functions as a virulence
factor in various pathogenic bacteria (18, 19). In *A. xylosoxidans*, the activity of the T6SS
contributes to the internalization in immune cells wherein it expresses cytotoxic effectors (12,
19). Some environmental isolates also contain phage regions encoding components of type IV
secretion systems (T4SS), and several genetic islands that may enhance survival in host-free
environments (10). In addition to secretion systems, motility is a major virulence factor in
pathogenic bacteria with *A. xylosoxidans* producing peritrichous flagella (20). Loss of
functional flagella is commonly associated with reduced virulence (21).

The genus *Achromobacter* is well-known for its inherent resistance to a broad range of
antimicrobial agents (22) and its ability to acquire resistance to commonly used treatments
for respiratory infections, such as colistin and carbapenems (22, 23). *A. xylosoxidans* encodes
at least 50 drug resistance – associated genes, including β-lactamases, efflux pumps, and
transposons (24). Seventeen different efflux pumps have been identified in *A. xylosoxidans,*
which may contribute to its intrinsic resistance (24). However, to date, only three of these
efflux pumps have been characterized. Notably, the observed susceptibility profiles often
poorly correlate with the presence of putative resistance genes (24). These findings
highlight the need to investigate the prevalence and mechanisms of antibiotic resistance
(AR) among *A. xylosoxidans* isolates. A powerful tool for the investigation of bacterial
genomics is supervised machine learning (ML). Supervised ML has been implemented to
identify the resistance profiles of multiple organisms based on genomic characteristics alone
(25). Usage of this technique therefore enables the prediction of antimicrobial resistance
and the identification of potential resistance genes across *A. xylosoxidans* isolates (26-29).

Biofilm formation can contribute significantly to antibiotic resistance, as the biofilm matrix
provides a protective environment that limits antibiotic penetration and promotes the survival
of persistent cells (30). Additionally, motility structures such as flagella may play a role in the
initial attachment and establishment of biofilms. *A. xylosoxidans* appeared to downregulate
its flagellar gene expression during biofilm formation (31).

In this work, we examined the genomic diversity of all available genome sequences of *A.*
xylosoxidans. Phylogenetic analysis was performed on a total of 178 genomes, including those
from NCBI and our recently collected clinical isolates from the NIH clinical center. Given the
critical role of secretion systems and motility in virulence, we analyzed the presence of all
secretion systems (types I to IX) and flagellar system in our NIH clinical isolates. Phylogenetic
analysis was further used to classify the T3SS family of *A. xylosoxidans*. We used our previously
collected whole genome sequences and susceptibility testing data from the clinical isolates to
predict AR among NCBI genomes using machine learning. The cytotoxicity, biofilm and motility
were assayed.

2. Methods

Data used in this work:

Raw sequence data from Bio Project ID PRJNA1148967, referred to as NIH data with 38
Achromobacter xylosoxidans, *A. denitrificans, A. ruhlandii, A. mucicolens and A. insolitus*
genomes. 197 genomes of *Achromobacter xylosoxidans* from NCBI
(https://www.ncbi.nlm.nih.gov/datasets/genome/?taxon=85698&assembly_level=1:3) as of
September 2024, referred as NCBI data.

Genome assembly

For all NIH isolates: Adapters were trimmed using Trimmomatic version V0.39 (32) with
default parameters, except the last 10 bases were removed. FastQC (v0.11.9) (33) was used
to visualize read quality, which showed an average base sequence quality greater than 28
across the reads. Genome assembly was performed using SPAdes (34) in careful mode. The
quality of the assemblies was evaluated using QUAST (v5.2.0) (35), which provided metrics
such as number of contigs, length, and N50.

Genome filtering

For NIH data, contigs shorter than 500 base pairs or with coverage below 20x were removed.
For NCBI data, genomes with more than 300 contigs were excluded, and contigs shorter than
500 base pairs were removed. Potential sample contamination was assessed using CheckM
(v1.2.3) (36). Genomes with contamination levels exceeding 5% or completeness below 90%
were excluded from subsequent analyses.

Core genome analysis

Assembled genomes were annotated with Prokka (1.14.6) (37) with the annotation from
assembly ASM1672882v1 (strain FDAARGOS_1091) used as a reference database. Core
genome analysis was performed using the Panaroo pipeline version 1.5.0 (38) with a protein

sequence identity threshold of 90%, gene length coverage cutoff of 75%, and a core genome
sample threshold of 0.95.

Phylogenetic analysis of *Achromobacter xylosoxidans*

Phylogenetic analysis of NIH and NCBI data was performed using kSNP4 (39). The optimal k-
mer size was determined using Kchooser. kSNP4 was run with the selected k-mer size, and a
parsimony tree was generated based on SNP loci present in at least 95% of the genomes. For
NIH data, in addition to kSNP4, RaxML (40) was used to construct the phylogenetic tree.
Briefly, core genes identified by Panaroo were used as input. RAxML v8.2.12 (40) was
utilized to build a rooted maximum likelihood tree using the general time-reversible (GTR)
substitution model with gamma correction to account for rate variation among sites. Since
the phylogenetic trees generated by kSNP4 and RAxML were identical, only one tree was
reported. The final tree was visualized using iTOL (41).

Population structure analysis

The genome-wide haplotype data were calculated as described previously (42). SNP calling
was performed for core genes identified by the panaroo pipeline, and imputation for
polymorphic sites with a missing frequency of less than <1% was conducted using BEAGLE
v.3.3.2 (43). This genome-wide haplotype dataset contained 144,290 SNPs, which was used
to define isolate populations and subpopulations based on haplotype profile similarity.
fineSTRUCTURE analysis (44) was performed with 200,000 iterations of both the burn-in and
Markov chain Monte Carlo (MCMC) method to cluster individuals based on the coancestry
matrix, as previously described (45). The results were visualized as a heat map, with each
cell representing the proportion of DNA a recipient receives from each donor.

Variant calling

The variants were called with Snippy v4.4.0 (https://github.com/tseemann/snippy), with
default option. NIH-016-2, collected first in the NIH-016 series, was used as a reference.

Secretion system analysis

Secretion systems and appendages were identified using MacSyFinder 2.1.3 (46). TXSScan
was run to search for all types of secretion systems and appendages, including T1SS, T2SS,
T3SS, T4SS, T5SS, T6SS, Flagellum, Tad pilus, and T9SS. CONJScan was run to detect
conjugative and mobilizable elements, which include T4SSs. The "ordered replicon" option
was used for our complete genome sequences.

For genomes in which the secretion systems were not found by MacSyFinder, we manually
looked up for key genes of the systems in the annotated genomes, as it is possible that the
segmentation of the sequences that feed into MacSyFinder interferes with the gene search.
These are also key genes used by MacSyFinder. A system was noted as absent when none of
the genes were found.

Phylogenetic analysis of T3SS

Amino acid sequences of SctN or its homologs were extracted from Uniprot and NCBI.
Multiple sequence alignment of SctN sequences was performed using MAFFT (47) with
default parameters. A rooted maximum likelihood tree was constructed using RAxML v8.2.12
(40) with the general time-reversible (GTR) substitution model and gamma correction for
rate variation among sites. The phylogenetic tree was visualized using iTOL (41).

Structural Modeling of YscN ATPase

The amino acid sequence of the T3SS ATPase from *A. xylosoxidans* was obtained from NCBI
with sequence ID of WP_020925211.1. Structure of YscN was predicted using AlphaFold via
the website: https://alphafoldserver.com/. The predicted structure of *Pseudomonas*
aeruginosa PA14 was downloaded from https://alphafold.ebi.ac.uk/entry/A0A2U2Y0I4 with
accession number of AF-A0A0H2Z9F1-F1. Structural alignment of the two proteins was
performed using Pymol (The PyMOL Molecular Graphics System, Version 3.0 Schrödinger,
LLC).

Biofilm

Bacterial cultures were initiated in glass culture tubes by inoculating bacteria from LB agar
plates into 1 mL of LB broth at a starting OD600 of 0.1. The cultures were then incubated
statically at 37°C for 3 days. On the third day, media was removed by pipetting, and biofilms
were washed with water and dried overnight at 37°C. The following day, a 1% crystal violet
solution was used to stain biofilms for 5 minutes, followed by additional washes with water
and overnight drying at 37°C. Biofilms were solubilized with 100% DMSO, and absorbance
was measured at 595 nm using a BioTek plate reader.

Infection of *Achromobacter* in cell lines

J774a.1 cells were seeded at a density of $6x10^4$ cells in 100 µL DMEM media in a tissue culture-
treated black flat-clear bottomed 96-well plate. On the day of infection, cells were washed
with PBS and incubated for one hour with OPTI-MEM media containing 200 ng/mL propidium

iodide (PI) and 1.45μM Hoechst 33342. *Achromobacter* inoculum was then prepared by
resuspending bacteria in OPTI-MEM. Cells were infected with the bacteria at multiplicity of
infection of 7 and spun down at 750 rpm for 5 minutes to synchronize infection. The 96-well
plate was then inserted into a stage incubator supplied with CO_2 and maintained at 37°C in a
Nikon AXR confocal fluorescent microscope. Three images of cells were taken at arbitrary
locations within each well and observed every hour for up to five hours. Cytotoxicity was
measured by the ratio of PI$^+$ signal to Hoechst$^+$ signal (dead cells/total cells) expressed in
percentage using ImageJ v 1.54g. PI$^+$ and Hoechst$^+$ cells were processed using background
subtraction, followed by gamma adjustment and median filtering. Then, Otsu method was
applied for thresholding, and cell counts were obtained using the Analyze Particles function.

Swimming motility in *Achromobacter* strains

Bacteria were inoculated and grown overnight in 1mL LB broth from single colonies on LB agar
plates. Overnight growth was measured by OD600 and concentrated to an OD600 of 10 in
100μL. 2μL of the resulting suspension was injected via pipette into 0.25% agarose 2%
tryptone plates. Plates were incubated at 37°C for 4 hours and imaged with a Chemidoc^{tm}
Imaging System using the stain free blot procedure with 3 second exposure time. The resulting
images were exported to ImageJ. Images were converted to an 8-bit format and a
brightness/contrast correction step was performed. The zone of motility was then selected
using the circle tool and resulting area in Pixels was exported. Pixel count was converted to
mm^2 using the formula: $mm^2 = (Pixels)/(5000\ Pixels\ /((3.5mm/2)^2*3.14159))$.

3. Results

Genomic diversity of *Achromobacter*

To examine the diversity among *A. xylosoxidans* isolates, we analyzed the whole-genome
sequencing data from 48 *Achromobacter* spp. isolates (including 34 *A. xylosoxidans*, and other
Achromobacter species collected at the NIH clinical center in the previous study (51) referred
to as NIH isolates) and 197 *A. xylosoxidans* genomes from NCBI (referred to as NCBI isolates).
Although our focus was on *A. xylosoxidans*, we included other *Achromobacter* species from
our NIH isolates as references to assess their clustering relative to *A. xylosoxidans*. The 197
NCBI isolates were predominantly from the United States (n = 51), Germany (n = 39), Russia
(n = 19), China (n = 12), and Australia (n = 11) (Figure 1). In this dataset, *A. xylosoxidans* isolates
were primarily collected from respiratory sources (e.g. cystic fibrosis patients), followed by
environmental samples (Figure 1A). Among the NIH isolates, the majority of samples were
collected from respiratory sources, with bloodstream isolates being the second most common
(Figure 1B).

From the NIH isolate sequence, genome assembly was performed and contigs greater than 500 bps with coverage of 20 or more were retained. CheckM analysis identified 10 genomes with contamination levels exceeding 5%, which were removed from further analysis (Table S1). This resulted in 38 genomes from the NIH isolate dataset, including 24 *A. xylosoxidans* isolates. For the NCBI genomes, those with more than 300 contigs were excluded, and contigs shorter than 500 bps were removed, resulting in 141 genomes. One genome was further excluded due to low genome completeness (less than 70%) (Table S1). This yielded a final dataset of 178 genomes for downstream analysis.

We first examined the core genome of these isolates. Using the Panaroo pipeline, we identified 3,387 core genes (present in at least 95% of the isolates), representing 53% of total genes in *A. xylosoxidans* (Table 1). The remaining 47% of genes were classified as accessory genes, highlighting the genomic plasticity of this species. Phylogenomic analysis of the core genomes of all 178 isolates revealed significant diversity, with isolates largely being nonclonal (Figure 2). NIH isolates were scattered among NCBI isolates, and clustering did not correlate with geographical locations or sample sources. A group of environmental isolates formed a distinct cluster (except for NIH-006-1), while others clustered with NIH clinical isolates, underscoring the complexity of *A. xylosoxidans*. Non-*A. xylosoxidans* isolates clustered together (except for one isolate NIH-021-1), suggesting the ability of this analysis to differentiate *Achromobacter* species. In co-ancestry analysis, there are a few cases with some level of shared ancestry from environmental samples. However, most isolates are unrelated to each other, suggesting that they are genetically diverse and likely evolve independently (Figure S1).

We also specifically looked at the phylogenetic analysis of the *A. xylosoxidans* genomes from the NIH isolates (Figure 2B). In this data, isolates collected from the same patients over time have the same middle number. For example, NIH-016-2 and NIH-016-3 were from the same patients; or NIH-010-3, NIH-010-4, NIH-010-6 were from the same patients. Phylogenetic analysis using two methods, kSNP or Panaroo followed by RaxML (see Methods section for details) yielded identical phylogenetic trees for the NIH isolates data, suggesting the robustness of the phylogenetic reconstruction. The analysis showed that isolates collected from the same patients clustered together, indicating genetic similarity and suggesting persistence and evolution within the same host over time. The only exception was NIH-005-4, which did not cluster with NIH-005-2 and NIH-005-3. NIH-005-4 was collected from sputum in 2014, whereas NIH-005-2 and NIH-005-3 were collected from the perirectal region and bronchoalveolar lavage, respectively, in 2016.

Antibiotic resistance in *Achromobacter*

Antibiotic susceptibility data of NIH isolates revealed a prevalence of multidrug-resistant *Achromobacter* spp. among the 12 antibiotic tested, including CZA (Ceftazidime/Avibactam), C/T (Ceftolozane/Tazobactam), IMR (Imipenem/Relebactam), MERO (Meropenem), AZT/ATM (Aztreonam), FEP (Cefepime), TAZ (Ceftazidime), CIP (Ciprofloxacin), IMI (Imipenem), LEVO (Levofloxacin), PT4/TZP (Piperacillin/Tazobactam), and SXT (Trimethoprim/Sulfamethoxazole) ((51), Figure 3A). Many AR genes were identified in the NIH isolates using the CARD database; however, these genes could not fully explain the AR in certain isolates (51). Two notable isolates, NIH-016-2 and NIH-016-3, were collected from the same patient. However, NIH-016-3 was resistant to 11 antibiotics (all antibiotic tested except SXT), while NIH-016-2 was resistant to only three. Antimicrobial Resistance analysis using CZ ID website (52) showed that NIH-016-2 and NIH-016-3 isolates carried similar sets of AR systems, including efflux pumps (AdeFGH, AxyXY-OprZ, MexAB-OprM, MuxABC–OpmB, MexGHI-OpmD, MexCD-OprJ, MexJK-OprM, MexPQ-OpmE), OXA-114h β-lactamase, AXC β-lactamase, and CatB. Thus, we re-sequenced both NIH-016-2 and NIH-016-3, as well as NIH-016-1, which has the same level of antibiotic resistance as NIH-016-3 and was collected after NIH-016-2. Using both short reads and long read sequences, we were able to assemble complete genomes of the three isolates. Then, variant analysis was done using NIH-016-2 as a reference (Table 3). Interestingly, the variant analysis identified a T>G single nucleotide polymorphism (SNP) at position 29 of AxyZ, resulting in a nonsynonymous mutation (V29G) in both NIH-016-3 and NIH-016-1. AxyZ is a negative regulator of AxyXY-OprZ RND efflux pump. This mutation is associated with overexpression of AxyXY-OprZ efflux pump, which may involve in resistance against multiple antibiotics including tobramycin, gentamicin, cefepime, ofloxacin, and levofloxacin (53). Thus, it is possible that the mutation is contributing to higher resistance to resistance to FEP and LEVO in NIH-016-3 and NIH-016-1.

Several other SNPs that may involve in antibiotic resistance were found. A few of them have been found to be involved in antibiotic resistance in other bacteria. A T>A SNP at position 695 of MutL in NIH-016-3 was identified. This mutation results in a nonsynonymous mutation from leucine to glycine (Leu232Gly). MutL, along with MutS and UrvD belong to mismatch repair system responsible for DNA repair. Mutations in MutL and MutS are associated with hypermutator phenotype and higher antibiotic resistance in *P. aeruginosa* (54). Similarly, MutS deletion has been associated with hypermutation and identified as an important adaptive mechanism of *A. xylosoxidans* during chronic infection (55). Thus, this mutation might be another potential contributor to higher resistance in NIH-016-03. A variant T>C at position 2228 in *mfd* gene which led to a change from leucine to proline (Leu743Pro) mutation was identified. This gene has been associated with increasing rate of mutagenesis and the development of antibiotic resistance (56). Additionally, C>T SNP leading to a change from arginine to cysteine (Arg198Cys) mutation in ABC transporter substrate-binding protein was identified. ABC transporters are essential in controlling the uptake and efflux of different substrates including antibiotics in a bacterial cell (57). Overall, there are at least three SNPs potentially impact AR of NIH-016s, that requires further study.

Secretion systems in *Achromobacter xylosoxidans* clinical isolates

Previous studies have shown that *A. xylosoxidans* can express multiple virulence factors,
including secretion systems and flagella (15, 23). We performed a bioinformatics analysis to
identify secretion systems and flagella in the 24 *A. xylosoxidans* NIH isolates. The reference
genome (FDAARGOS_1091) and two additional clinical isolates (GN050 and GN008) were
included for comparison (Table 3). All isolates possessed T1SS, T2SS, T5SS, and bacterial Tight
adherence Secretion System (TadSS). T6SS was present in all isolates except NIH-021-1, and
T4SS was present in all isolates except NIH-009-1. Interestingly, although a number of isolates
were collected from the same patients over time, most of them still carry flagella genes and
T3SS. Only one isolate, NIH-023-1 doesn't encode for either the flagellar system or the T3SS.

The type III secretion system is an important virulence mechanism contributing to cytotoxicity
in *A. xylosoxidans* (12, 15). Thus, we examined the T3SS in *A. xylosoxidans* more closely.
Phylogenetic analysis showed that the T3SS in *A. xylosoxidans* belongs to the Ysc T3SS family
(Figure 4A), which includes T3SS from other pathogens such as *P. aeruginosa*, *Yersinia*
pseudotuberculosis, and *Bordetella pertussis*. Structural modeling of the APTase YscN from *A.*
xylosoxidans showed high similarity to that of *P. aeruginosa* PA14, with a root-mean-square
deviation (RMSD) of 0.488 A$^{\circ}$ (Figure 4B). This structural similarity further supports the
classification of *A. xylosoxidans* T3SS within the Ysc family.

Cytotoxicity in *Achromobacter*

Previous studies have shown that T3SS expression correlates with cytotoxicity in infected host
cells (12, 15). To assess the cytotoxicity of the NIH isolates, we infected macrophage J774.a.1
with each isolate and measured cell death over time. We included two clinical isolates, GN050
and GN008, which were previously reported to exhibit differential cytotoxicity, as controls
(15). The result showed that the 24 isolates have variable cytotoxicity levels with NIH-018-1
being the most cytotoxic, exceeding that of GN050 (Figure 5). Among clinical isolates no
measurable difference was detected between the isolate source, and no clear pattern was
apparent in isolates collected at subsequent time points. Among the isolates collected from
the same patients, NIH-010 and NIH-016 had minal cytotoxicity, while NIH-025 are cytotoxic.
In NIH-018 series, the bacteria seemed to be more virulence over time. Most cytotoxic isolates
possess T3SS and AxoU (Table 3), with one exception. Isolate NIH-019-1 has an intermediate
cytotoxicity level, but it doesn't encode AxoU, although it has T3SS. This may suggest the
presence of (an) uncharacterized effector(s).

Biofilm

The capacity of a bacterium to form biofilms is a strong indicator of virulence and can be tied
to the expression of other virulence factors (58) (59) (60). Additionally, biofilm formation
improves the resistance of the organism to antibiotic and environmental pressures. We
therefore set out to examine biofilm formation in static culture of *A xylosoxidans*. *P.*
aeruginosa PA01 Δ*wspF*Δ*pelA*Δ*pslBCD* was used as negative control and PA01 Δ*wspF* as a

positive control (61, 62) for biofilm formation. GN050 showed robust biofilm formation
compared to the control (Figure 6). Most NIH isolates demonstrated a low level of biofilm
production. Only NIH-016-1, NIH-021-1 and NIH-025s displayed substantial biofilm forming
capabilities over the negative control. These isolates were from sputum (NIH-016-1), wound
(NIH-021-1), blood or lung (NIH-025s), suggesting that the source of isolation does not
determine whether the bacteria can form biofilm. The intra-host series, NIH-010, NIH-016 and
NIH-018 do show a trend of increasing biofilm production although it's not statistically
significant from the *P. aeruginosa* negative control.

***Achromobacter* Flagellar Motility**

Flagellar motility is an important virulence factor which the bacteria employ to move to the
site of infection and is often coordinated with the expression of other virulence genes (63-65).
During prolonged infection, mutation in flagellar genes can accumulate resulting in an overall
loss of flagellar production (66). We decided to investigate the flagellar motility of our *A.*
xylosoxidans clinical isolates to determine if our strains demonstrate any changes in motility.
Our results demonstrated substantial differences in flagellar motility between strains even
from isolates collected from the same patients (Figure 7A). The most motile isolates are NIH-
003-1, NIH-005-4, NIH-014-1, NIH-016-2, NIH-018-3, and NIH-026-1. Samples isolated from
respiratory systems showed a greater tendency to lose motility compared to those from skin
or wound infections (Table 3). Among all 14 respiratory isolates (NIH-005-3, NIH005-4,
NIH009-1, NIH010-3, NIH-010-4, NIH-014-1, NIH016-2, NIH016-3, NIH018-1, NIH018-2,
NIH019-1, NIH022-1, NIH023-1, and NIH025-6) 8 were identified as non-motile representing
57.1% of the group. Additionally, among all non-motile isolates, 80% were isolated from a
respiratory source. Comparing strains isolated at different times from the same patient, NIH-
005, NIH-010, and NIH-016 series showed a loss of motility whereas NIH-018 series showed
reduced motility. In NIH-005 series, isolate 2 and 3 were collected in the same year, two years
after isolate 4. The isolates 2 and 3 don't move while isolate 4 demonstrates high motility
among all isolates. This also supports the phylogeny finding that isolate 4 is different from
isolates 2 and 3. In isolates NIH-010 series, isolate 6 was more motile than isolates 3 and 4,
which were isolated after isolate 6. In series NIH-018, NIH-018-3 is one of the best swimmers.
The motility reduced about 50% after one and two years in the subsequent isolates, but they
are still motile. These observations indicate a preference for a reduction of flagellar function
in the respiratory environment.

The NIH-016 series demonstrated some of the most substantial differences in motility
between isolates. We investigated the genomic differences present among genomes of NIH-
016 series. Multiple sequence alignment between known flagellar genes in NIH-016-2
compared to the genome of NIH-016-3 showed near perfect identity between the isolates
except for two identified point mutations in *fliO* and *flhF* (supplemental table 1). The point
mutation in *fliO* when translated results in a synonymous mutation. However, a T>G single
nucleotide polymorphism was identified in *flhF* which resulted in a Y330H mutation FlhF

(Figure 7B). This mutation also exists in NIH-016-1 which was collected after NIH-016-3. FlhF
is a GTPase known to be important for the initial insertion of flagellar components in the
membrane of the bacteria. Deletion or mutations of key regions of this protein are known to
result in non-flagellated bacteria and thus a loss of motility (67, 68).

Characterization of the protein for domains by InterProScan (69) revealed a predicted SRP54-
type protein, GTPase domain which spans from 329 to 520. SRP54-type proteins represent
signal recognition particles which bind GTP and are important for trafficking to the plasma
membrane (68, 70). This observation is in line with the predicted function of FlhF as a GTP
binding protein and its function in nucleating flagellar assembly in other bacteria (71, 72).
Prediction of the FlhF protein overall via modeling in alphafold (73) revealed four distinct
structures. A structure ranging from 268 to 520 comprised a GTPase domain. Alphafold
modeling of NIH-016-2 FlhF and NIH-016-3 FlhF and subsequent alignment revealed a
structural abnormality in NIH16-3's FlhF which resulted in a predicted displacement of a valine
at position 329 (Figure 7C). FlhF is known to homodimerize following binding to GTP (70) and
therefore we generated an additional model to visualize plausible homodimer structures of
NIH-0016-2 FlhF. Y330 was observed on the outside of the predicted model, indicating that
the mutation may not be causing any structural disruptions directly responsible for
dimerization (Figure 7D). An alternative hypothesis is the mutation may be disrupting the GTP
binding domain. An alignment of the predicted *Achromobacter* FlhF structure to the known
crystal structure of FlhF in *B. subtilis* showed a R^2 of 2.082, demonstrating they are structurally
similar. The known location of the GTP binding domain of *B. subtilis* FlhF is in close proximity
to the β-sheet containing the Y330H mutation (7E). Using observed differences in flagellar
motility we were able to identify a mutation in NIH-016-3 and NIH-016-1 which may be
contributing to a loss of flagellar motility possibly by the disruption of GTP binding. Overall,
these results demonstrate the varying dynamics of flagellar motility of *Ax* clinical isolates and
a loss of flagella activity during chronic infection.

4. Figures and tables

Figure 1: Geographic locations and collection source of samples used in this study.
(A) The number of isolates collected in different countries around the world. (B) The number of isolates collected from different sources.

Figure 3: Frequency of antibiotic resistance in NIH isolates. (A) Antibiotic resistance in NIH clinical isolates. The number of antibiotics out of 11 tested that each isolate was resistant to was plotted. (B) Percentage of isolates resistance to each antibiotic was shown.

465 **Figure 4: The type III secretion phylogeny and structure.** A. Phylogenetic tree of T3SS families.
466 The tree was prepared using aligned amino acid sequence of SctN and its homologs. **B**
467 Predicted structure of YscN from *A. xylosoxidans*. **C.** Predicted structures of *A. xylosoxidans*
468 YscN (Orange color)and *Pseudomonas aeruginosa* PA14 PscN (blue color) superimposed.

A.

B.   C.

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Figure 5: Cytotoxicity of *A. xylosoxidans* clinical isolates. (A) Cell death of macrophage J774a.1 cells infected with selected clinical isolates of *A. xylosoxidans* over time. (B) Cell death of macrophage J774a.1 cells infected with different clinical isolates of *A. xylosoxidans* at 5hai. Percent cell death - the ratio between dead cells and total cells - were shown. Different colored dots are for different biological replicates. Each isolate was experimented for at least two biological replicates, three technical replicate each time.

Figure 6: Biofilm formation by _A. xylosoxidans_ clinical isolates. PA01
$\Delta wspF\Delta pelA\Delta pslBCD$ was used as a negative control. PA01 $\Delta wspF$ was used as a positive
control. Different color dots indicate different biological replicates. The experiment was
done in at least three technical replicates. Kruskal-Wallis test was done followed by Dunn's
multiple comparisons to correct for multiple comparison. **** P-val <0.0001, *** P-val
<0.0002, ** P-val <0.002, *P-val <0.0332.

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Figure 7: *A. xylosoxidans* **motility.** (A) *A. xylosoxidans* demonstrates substantial differences in flagellar motility among clinical isolates. (B) NIH-016-3 FlhF contains a point mutation resulting in a Y330H. (C) Mutation Y330H results in a predicted displacement of V229 (Purple) as determined by modeling in Alphafold (Blue: NIH-016-2 FlhF, Red: NIH-016-3 FlhF). (D) Alignment of monomeric NIH-016-2 FlhF (Blue) to homodimerized NIH-016-2 FlhF (Yellow). (E) Alignment of NIH-016-2 (Blue) to *B. subtilis* FlhF (Green) with bound GTP (Red).

Figure S1: Co-ancestry analysis based on fineSTRUCTURE. The color of each cell in the matrix indicates the expected number of chunks imported from a donor genome (column) into a recipient genome (row). The sample source (A) and top six locations (B) were color coded next to the sample ID.

Figure S2: Correlation of cytotoxicity, biofilm and motility. Correlation between cytotoxicity
and motility (A), cytotoxicity and biofilm (B) and motility and biofilm (C), and biofilm and
antibiotic resistance (the number of antibiotic each isolate resistant to)

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Table 1: Core genes in *Achromobacter xylosoxidans*
* Rare genes that occurs in less than 15% of isolates re not
counted in the total genes to avoid underestimate the core gene
for each genome.

Table 1A. core genes from 178 NIH and NCBI genomes	counts
Core genes (99% <= strains <= 100%)	2074
Soft core genes (95% <= strains < 99%)	1313
Shell genes (15% <= strains < 95%)	3066
Cloud genes (0% <= strains < 15%)	23298
Total genes *(15% <= strains <= 100%)	6453

Table 1B. core genes from 24 NCBI genomes	counts
Core genes (99% <= strains <= 100%)	4149
Soft core genes (95% <= strains < 99%)	588
Shell genes (15% <= strains < 95%)	2420
Cloud genes (0% <= strains < 15%)	3020
Total genes* (15% <= strains <= 100%)	10177

Table 2: Summary of secretion system, motility, cytotoxicity, and biofilm of *A. xylosoxidans* isolates. Three intra-host isolate series were highlighted in different colors. NIH-016-1 and NIH-018-3 were included in the table for the analysis of intra-host samples, while the genome sequencing were omitted in the phylogenetic analysis due to contamination issue.

*motility in mm^2

#cytotoxicity: percent cell death

$ biofilm: Absorbance 595nm of crystal violet stained biofilm

BAL: Bronchoalveolar lavage, AXL: Autopsy of the lung

Isoates	T1SS	T2SS	T3SS	T4SS	T5SS	T6SS	Tad	flagella	motility*	cytotoxicity#	presence of AxoU	Biofilm$	Total antibiotic resistance	collection date	Source##
NIH-003-1	yes	yes	yes	yes	yes	yes	yes	yes	16.9	5.9	yes	0.39	2	12/17/17	Wound
NIH-005-4	yes	yes	yes	yes	yes	yes	yes	yes	17.2	9.7	yes	0.34	2	10/10/14	Sputum
NIH-005-3	yes	yes	yes	yes	yes	yes	yes	yes	1.9	5.6	yes	0.11	9	10/30/16	BAL
NIH-005-2	yes	yes	yes	yes	yes	yes	yes	yes	2.1	3.9	yes	0.12	10	11/10/16	Perirectal
NIH-009-1	yes	yes	yes	no	yes	yes	yes	yes	9.5	2.8	yes	0.29	2	8/29/16	BAL
NIH-010-6	yes	yes	yes	yes	yes	yes	yes	yes	6.7	3.5	yes	0.20	5	9/4/12	Skin lesion
NIH-010-4	yes	yes	yes	yes	yes	yes	yes	yes	0.6	9.8	yes	0.19	5	7/26/13	BAL
NIH-010-3	yes	yes	yes	yes	yes	yes	yes	yes	0.7	1.8	yes	0.31	7	7/5/14	Sputum
NIH-014-1	yes	yes	yes	yes	yes	yes	yes	yes	15.1	3.1	yes	0.31	3	11/16/15	Autopsy lung
NIH-016-2	yes	yes	yes	yes	yes	yes	yes	yes	13.5	7.5	yes	0.31	3	7/15/14	Sputum
NIH-016-3	yes	yes	yes	yes	yes	yes	yes	yes	0.7	2.5	yes	0.34	11	2/11/15	Sputum
NIH-016-1	yes	yes	yes	yes	yes	yes	yes	yes	0.6805	5.4	yes	0.52	11	2/24/16	Sputum
NIH-017-1	yes	yes	yes	yes	yes	yes	yes	yes	0.6	5.7	yes	0.30	2	1/8/15	Nephrostomy urine
NIH-018-3	yes	yes	yes	yes	yes	yes	yes	yes	17.4	15.5	Yes	0.26	6	12/13/12	BAL
NIH-018-2	yes	yes	yes	yes	yes	yes	yes	yes	8.4	26.8	yes	0.21	6	11/8/13	Sputum
NIH-018-1	yes	yes	yes	yes	yes	yes	yes	yes	9.3	45.3	yes	0.47	6	1/22/14	Sputum
NIH-019-1	yes	yes	yes	yes	yes	yes	yes	yes	0.8	11.0	no	0.22	7	4/2/14	Sinus
NIH-021-1	yes	yes	yes	yes	yes	no	yes	yes	5.0	4.1	no	0.47	2	10/22/12	wound

NIH-022-1	yes	yes	yes	yes	yes	yes	yes	yes	2.7	11.7	yes	0.25	2	12/7/12	Sputum
NIH-023-1	yes	yes	no	yes	yes	yes	yes	no	0.5	4.9	no	0.31	3	3/22/13	BAL
NIH-025-1	yes	yes	yes	yes	yes	yes	yes	yes	8.9	29.9	yes	0.38	8	11/12/13	Blood
NIH-025-3	yes	yes	yes	yes	yes	yes	yes	yes	7.1	20.1	yes	0.32	9	11/20/13	Blood
NIH-025-4	yes	yes	yes	yes	yes	yes	yes	yes	5.1	28.8	yes	0.35	9	11/22/13	axl
NIH-025-5	yes	yes	yes	yes	yes	yes	yes	yes	5.5	11.8	yes	0.38	8	11/10/13	Blood
NIH-025-6	yes	yes	yes	yes	yes	yes	yes	yes	7.1	29.6	yes	0.51	9	11/9/13	BAL
NIH-026-1	yes	yes	yes	yes	yes	yes	yes	yes	20.6	4.7	yes	0.32	2	5/1/16	Wound (l heel)
GCA_016728825.1 (reference)	yes	yes	yes	yes	yes	yes	yes	yes	ND	ND	yes	ND	ND	4/30/18	Soil
GCA_013282235.1 (GN050)	yes	yes	yes	yes	yes	yes	yes	yes	8.0	36.7	yes	0.74	ND	7/12/10	Ear effusion
GCA_013282255.1 (GN008)	yes	yes	yes	yes	yes	yes	yes	yes	0.4	3.6	yes	0.35	ND	6/29/10	Cystic fibrosis sputum

Table 3: Changes of the genome between NIH-016-2 and NIH-016-1, NIH-016-3. NIH-016-2 was used as a reference. The mutations listed are the common mutations between NIH-016-3 and NIH-016-1, compared to NIH-016-2.

CHROM	POS	REF	ALT	EFFECT	LOCUS_TAG	PRODUCT
CP188777	5147082	G	GC	frameshift_variant c.1181dupC p.Pro395fs	ACP6EQ_23 745	DUF3772 domain-containing protein
CP188777	5749106	G	GC	frameshift_variant c.1721dupG p.Ala575fs	ACP6EQ_26 625	autotransporter domain-containing protein
CP188777	6041085	G	GC	frameshift_variant c.25dupC p.Leu9fs	ACP6EQ_27 890	response regulator
CP188777	2069911	A	AC	frameshift_variant c.278dupC p.His94fs	ACP6EQ_09 695	glycolate oxidase subunit GlcE
CP188777	6486625	CG	C	frameshift_variant c.421delG p.Glu141fs	ACP6EQ_29 965	HIT family protein
CP188777	6419572	A	ACG	frameshift_variant c.481_482dupGC p.Tyr162fs	ACP6EQ_29 660	dTDP-4-amino-4,6-dideoxygalactose transaminase
CP188777	4951892	G	GC	frameshift_variant c.73dupG p.Ala25fs	ACP6EQ_22 700	PBSX family phage terminase large subunit
CP188777	4560728	G	GC	frameshift_variant c.839dupG p.Val281fs	ACP6EQ_20 870	molybdopterin cofactor-binding domain-containing protein
CP188777	3770678	A	G	missense_variant c.1001T>C p.Val334Ala	ACP6EQ_17 480	D-cysteine desulfhydrase family protein
CP188777	6213780	T	C	missense_variant c.1043T>C p.Val348Ala	ACP6EQ_28 685	30S ribosomal protein S12

						methylthiotransferase RimO
CP188777	6348240	C	T	missense_variant c.104G>A p.Gly35Asp	ACP6EQ_29305	alpha/beta fold hydrolase
CP188777	306544	C	T	missense_variant c.106C>T p.Arg36Cys	ACP6EQ_01455	ornithine cyclodeaminase
CP188777	5614531	A	G	missense_variant c.10T>C p.Cys4Arg	ACP6EQ_26035	Bug family tripartite tricarboxylate transporter substrate binding protein
CP188777	6585839	C	T	missense_variant c.1195G>A p.Glu399Lys	ACP6EQ_30490	catalase
CP188777	2676682	C	T	missense_variant c.1264G>A p.Val422Met	ACP6EQ_12500	heavy metal translocating P-type ATPase
CP188777	785311	A	G	missense_variant c.1313T>C p.Leu438Pro	ACP6EQ_03715	ribonuclease catalytic domain-containing protein
CP188777	2395874	A	G	missense_variant c.133A>G p.Thr45Ala	ACP6EQ_11165	DUF5339 family protein
CP188777	5525409	T	C	missense_variant c.1396A>G p.Thr466Ala	ACP6EQ_25620	TonB-dependent siderophore receptor
CP188777	2107651	C	T	missense_variant c.146C>T p.Ser49Leu	ACP6EQ_09910	aromatic amino acid DMT transporter YddG
CP188777	5968107	A	G	missense_variant c.179A>G p.Asn60Ser	ACP6EQ_27535	prolyl aminopeptidase

CP188777	4914304	T	C	missense_variant c.1876A>G p.Thr626Ala	ACP6EQ_22 475	TonB-dependent receptor
CP188777	1620551	A	G	missense_variant c.2228T>C p.Leu743Pro	ACP6EQ_07 680	transcription-repair coupling factor
CP188777	4918789	T	C	missense_variant c.230T>C p.Val77Ala	ACP6EQ_22 490	response regulator transcription factor
CP188777	3110891	T	C	missense_variant c.278A>G p.Tyr93Cys	ACP6EQ_14 500	Bug family tripartite tricarboxylate transporter substrate binding protein
CP188777	2194430	A	G	missense_variant c.299T>C p.Val100Ala	ACP6EQ_10 305	DUF3053 domain-containing protein
CP188777	1117935	C	T	missense_variant c.32G>A p.Gly11Asp	ACP6EQ_05 315	polynucleotide adenylyltransferase PcnB
CP188777	5368021	A	G	missense_variant c.358A>G p.Thr120Ala	ACP6EQ_24 820	protein translocase subunit SecD
CP188777	3172298	A	G	missense_variant c.415T>C p.Cys139Arg	ACP6EQ_14 805	GMC family oxidoreductase
CP188777	36790	A	G	missense_variant c.41A>G p.Asp14Gly	ACP6EQ_00 165	AraC family transcriptional regulator
CP188777	3111709	A	G	missense_variant c.422A>G p.Tyr141Cys	ACP6EQ_14 505	LysR family transcriptional regulator
CP188777	5994882	T	C	missense_variant c.439A>G p.Ser147Gly	ACP6EQ_27 665	RNA polymerase sigma factor

CP188777	896554	T	C	missense_variant c.443T>C p.Val148Ala	ACP6EQ_04 250	MFS transporter
CP188777	2516449	C	T	missense_variant c.451C>T p.Arg151Cys	ACP6EQ_11 765	class I SAM-dependent DNA methyltransferase
CP188777	3793425	AG	GC	missense_variant c.46_47delCTinsGC p.Leu16Ala	ACP6EQ_17 565	cystathionine gamma-synthase family protein
CP188777	5023271	T	C	missense_variant c.464A>G p.Tyr155Cys	ACP6EQ_23 155	4-hydroxythreonine-4-phosphate dehydrogenase PdxA
CP188777	444497	A	G	missense_variant c.472A>G p.Ser158Gly	ACP6EQ_02 080	Bug family tripartite tricarboxylate transporter substrate binding protein
CP188777	1828128	A	G	missense_variant c.4874A>G p.Tyr1625Cys	ACP6EQ_08 585	hemagglutinin repeat-containing protein
CP188777	1381223	A	G	missense_variant c.511A>G p.Thr171Ala	ACP6EQ_06 610	glycosyltransferase family 9 protein
CP188777	2494682	G	A	missense_variant c.553G>A p.Gly185Ser	ACP6EQ_11 675	copper-containing nitrite reductase
CP188777	3444638	G	A	missense_variant c.592C>T p.Arg198Cys	ACP6EQ_16 040	ABC transporter substrate-binding protein
CP188777	6398223	T	C	missense_variant c.688T>C p.Tyr230His	ACP6EQ_29 565	N-acetyl sugar amidotransferase

CP188777	1123581	A	T	missense_variant c.695T>A p.Leu232Gln	ACP6EQ_05 345	DNA mismatch repair endonuclease MutL
CP188777	1351955	T	G	missense_variant c.700A>C p.Thr234Pro	ACP6EQ_06 450	tyrosine-type recombinase/in tegrase
CP188777	755449	C	G	missense_variant c.700G>C p.Gly234Arg	ACP6EQ_03 575	DUF342 domain-containing protein
CP188777	642598	C	T	missense_variant c.733G>A p.Glu245Lys	ACP6EQ_03 095	bifunctional hydroxymethylp yrimidine kinase/phosph omethylpyrimid ine kinase
CP188777	497968	A	G	missense_variant c.74T>C p.Val25Ala	ACP6EQ_02 315	c-type cytochrome
CP188777	5144580	A	G	missense_variant c.74T>C p.Val25Ala	ACP6EQ_23 735	acetyl-CoA hydrolase/trans ferase family protein
CP188777	5074135	A	G	missense_variant c.818T>C p.Val273Ala	ACP6EQ_23 405	ABC transporter substrate-binding protein
CP188777	1704773	C	T	missense_variant c.848G>A p.Arg283His	ACP6EQ_08 075	Bug family tripartite tricarboxylate transporter substrate binding protein
CP188777	1587091	T	G	missense_variant c.86T>G p.Val29Gly	ACP6EQ_07 535	multidrug efflux transcriptional repressor AxyZ
CP188777	2510615	G	C	missense_variant c.88G>C p.Val30Leu	ACP6EQ_11 745	SDR family oxidoreductase

CP188777	1924181	G	A	missense_variant c.98G>A p.Gly33Asp	ACP6EQ_09 000	1,2-phenylacetyl-CoA epoxidase subunit PaaA

529 **Table S1: Assembly summary of NIH isolates**

Isolates	Contigs	Total length (>= 1000 bp)	N50	GC%	final contigs	Complete-ness	Contami-nation
NIH-010-1	154	6778138	81366	67.73	154	98.28	20.69
NIH-025-2	217	6917134	64404	67.59	213	98.28	10.34
NIH-010-2	213	6846762	60225	67.83	213	99.07	9.03
NIH-011-1	146	6714695	95958	67.77	146	99.53	7.94
NIH-018-3	141	6627574	94107	67.67	140	98.91	7.48
NIH-010-5	228	6582611	65494	67.7	202	98.28	6.9
NIH-027-1	157	7057208	92623	67.21	155	99.53	6.72
NIH-025-7	122	6736063	115055	67.59	122	99.07	6.23
NIH-013-1	264	7399130	58598	67.02	228	96.6	5.92
NIH-016-1	153	6938110	78345	67.68	152	99.24	5.61
NIH-015-1	219	7317467	69494	67.28	216	98.83	5.14
NIH-025-3	196	6971644	74558	67.55	196	99.3	4.21
NIH-016-3	162	6722923	73552	67.69	160	98.83	3.8
NIH-010-6	256	6589334	50077	67.69	254	99.51	3.27
NIH-010-4	183	6479449	67420	67.86	181	99.07	2.34
NIH-025-5	205	6816549	58213	67.6	201	98.95	2.02
NIH-016-2	97	6559927	124621	67.7	96	99.07	0.93
NIH-019-1	206	6967103	77893	68.34	203	99.53	0.93
NIH-022-1	129	6820766	102549	67.49	129	99.53	0.93
NIH-023-1	177	6736346	88615	67.51	175	99.53	0.93
NIH-003-1	122	6878569	94819	67.45	122	99.53	0.47
NIH-005-2	141	6466110	94875	67.79	141	99.53	0.47
NIH-005-3	106	6463122	116154	67.78	105	99.53	0.47
NIH-005-4	121	6727683	100896	67.65	119	99.53	0.47
NIH-009-1	75	6394470	194428	67.93	75	99.53	0.47
NIH-010-3	118	6438604	106660	67.83	118	99.07	0.47
NIH-014-1	97	6725191	135687	67.58	96	99.53	0.47
NIH-017-1	108	6968984	120623	67.49	107	99.3	0.47
NIH-018-1	143	6396229	81941	67.72	143	99.07	0.47
NIH-018-2	71	6407776	164746	67.72	71	99.53	0.47
NIH-025-1	138	6742034	90224	67.63	138	99.3	0.47
NIH-025-4	184	6682543	67033	67.58	184	98.6	0.47
NIH-025-6	181	6608399	67916	67.63	181	99.14	0.47
NIH-026-1	82	6375104	159180	67.87	81	99.07	0.47

| | NIH-021-1 | 118 | 6330035 | 97448 | 67.71 | 118 | 99.53 | 0 |

531 **Table S2: Assembly summary of NCBI isolates**

	Isolates	Completeness	Contamination
1	GCA_001641525_1	100	2.57
2	GCA_000165835_1	99.53	0.03
3	GCA_000508285_1	99.53	0.58
4	GCA_000967095_1	99.53	0.93
5	GCA_001038045_1	99.53	0.47
6	GCA_001276315_1	99.53	1.05
7	GCA_001457475_1	99.53	0.58
8	GCA_001471535_2	99.53	2.73
9	GCA_001541435_1	99.53	0.47
10	GCA_001558755_2	99.53	0
11	GCA_001559195_2	99.53	0.47
12	GCA_001566985_1	99.53	0.7
13	GCA_001598595_1	99.53	0.58
14	GCA_001654705_1	99.53	0.47
15	GCA_001707485_1	99.53	0
16	GCA_001810925_1	99.53	0.47
17	GCA_001957055_1	99.53	0.47
18	GCA_001957065_1	99.53	0.93
19	GCA_002083165_2	99.53	0.93
20	GCA_003031105_1	99.53	0.47
21	GCA_003177325_1	99.53	0.58
22	GCA_003293615_1	99.53	0.47
23	GCA_003503255_1	99.53	0
24	GCA_008432465_1	99.53	0.47
25	GCA_008631215_1	99.53	0.93
26	GCA_009363015_1	99.53	0.47
27	GCA_013282235_1	99.53	0.47
28	GCA_013282255_1	99.53	0.47
29	GCA_013343135_1	99.53	0.58
30	GCA_013413495_1	99.53	0.58
31	GCA_014353695_1	99.53	0
32	GCA_014529725_1	99.53	0
33	GCA_016403305_1	99.53	0.47
34	GCA_016599195_1	99.53	0.47
35	GCA_017745625_1	99.53	0
36	GCA_022320945_1	99.53	0.47
37	GCA_022320975_1	99.53	0.47
38	GCA_022321005_1	99.53	0.47

39	GCA_022321065_1	99.53	0.47
40	GCA_022321165_1	99.53	0.47
41	GCA_022321185_1	99.53	0.58
42	GCA_022321275_1	99.53	0.47
43	GCA_022321345_1	99.53	0.47
44	GCA_022325135_1	99.53	0.47
45	GCA_022325165_1	99.53	0.47
46	GCA_022325205_1	99.53	0.47
47	GCA_022444285_1	99.53	0.58
48	GCA_022488545_1	99.53	0.58
49	GCA_022524565_1	99.53	0.47
50	GCA_022524575_1	99.53	0.93
51	GCA_022524705_1	99.53	0.93
52	GCA_022524775_1	99.53	0.93
53	GCA_022524865_1	99.53	0.93
54	GCA_022525145_1	99.53	0.53
55	GCA_022525265_1	99.53	0.93
56	GCA_022525445_1	99.53	0.47
57	GCA_022976495_1	99.53	0.47
58	GCA_022976515_1	99.53	0.47
59	GCA_023703405_1	99.53	0.58
60	GCA_025426095_1	99.53	0.47
61	GCA_027529565_1	99.53	0.47
62	GCA_027529585_1	99.53	0.47
63	GCA_027529805_1	99.53	0.47
64	GCA_027529815_1	99.53	0.7
65	GCA_027922385_1	99.53	0.47
66	GCA_028462445_1	99.53	0.47
67	GCA_031994865_1	99.53	0.47
68	GCA_032703345_1	99.53	0.47
69	GCA_033096405_1	99.53	0.93
70	GCA_034424885_1	99.53	0.58
71	GCA_034424905_1	99.53	0.58
72	GCA_034424925_1	99.53	0.58
73	GCA_036035015_1	99.53	0.47
74	GCA_038277405_1	99.53	0
75	GCA_039595975_1	99.53	0.47
76	GCA_039674625_1	99.53	0.47
77	GCA_040438975_1	99.53	0.93
78	GCA_900009115_1	99.53	0.47
79	GCA_900009125_1	99.53	1.4
80	GCA_900010105_1	99.53	1.4

81	GCA_900475575_1	99.53	0.58
82	GCA_902363095_1	99.53	0.47
83	GCA_904067195_1	99.53	0.78
84	GCA_904067205_1	99.53	0.47
85	GCA_904067255_1	99.53	0.47
86	GCA_904067265_1	99.53	0.47
87	GCA_904067315_1	99.53	0.47
88	GCA_904067355_1	99.53	0.47
89	GCA_904067375_1	99.53	0.47
90	GCA_904067385_1	99.53	0.47
91	GCA_963528385_1	99.53	0.27
92	GCA_036527295_1	99.47	0.47
93	GCA_014490035_1	99.44	0
94	GCA_001299195_1	99.38	0
95	GCA_001299215_1	99.38	0.47
96	GCA_001558915_2	99.38	0.12
97	GCA_016027035_1	99.38	0.58
98	GCA_016728825_1	99.38	0.47
99	GCA_022870085_1	99.38	0.58
100	GCA_001299115_1	99.3	0.47
101	GCA_001299295_1	99.3	0.47
102	GCA_002192695_1	99.3	0
103	GCA_022321095_1	99.3	0.47
104	GCA_022325265_1	99.3	0.47
105	GCA_001299175_1	99.14	0.47
106	GCA_000517225_1	99.07	0.58
107	GCA_001051055_1	99.07	0
108	GCA_001298915_1	99.07	0.47
109	GCA_001957045_1	99.07	0.7
110	GCA_034078005_1	99.07	0.47
111	GCA_038024235_1	99.03	0
112	GCA_001299155_1	98.91	0.47
113	GCA_001815425_1	98.88	0.47
114	GCA_001299235_1	98.87	0.47
115	GCA_001298895_1	98.86	0.47
116	GCA_001299055_1	98.83	0.47
117	GCA_029837755_1	98.75	0.47
118	GCA_001299095_1	98.71	0.47
119	GCA_001298935_1	98.68	0.47
120	GCA_001299015_1	98.68	0.47
121	GCA_001299035_1	98.68	0.58
122	GCA_001299265_1	98.61	0.47

123	GCA_001299415_1	98.6	0.93
124	GCA_001813885_1	98.6	0.47
125	GCA_002294075_1	98.6	0.47
126	GCA_029837835_1	98.6	0.47
127	GCA_001298795_1	98.12	0
128	GCA_001299335_1	97.27	0.93
129	GCA_038024685_1	97.16	0
130	GCA_038024655_1	96.74	0.47
131	GCA_038024675_1	96.1	0.7
132	GCA_038024665_1	96.05	0.47
133	GCA_002453655_1	95.5	0.93
134	GCA_038024245_1	95.18	0.47
135	GCA_038024645_1	94.52	0
136	GCA_000186185_1	94.07	1.4
137	GCA_038049895_1	92.4	0
138	GCA_038049885_1	85.98	0.55
139	GCA_038049905_1	80.12	0.96
140	GCA_946222735_1	70.53	0
141	GCA_040396375_1	65.86	0

Table S3: Antibiotic susceptibility testing for NIH clinical isolates with IMI and IMR. R: resistance. S: susceptible. I: intermediate resistance.

De-ID	Source	IMI MIC (mg/L)	interpretation	IMR MIC (mg/L)	Interpretation
NIH-003-1	Wound	<=1	S	2	S
NIH-005-2	Perirectal	2	S	4	I^
NIH-005-3	BAL	2	S	4	I^
NIH-005-4	Sputum	<=1	S	2	S
NIH-009-1	BAL	4	S	4	I^
NIH-010-1	Sputum	2	S	4	I^
NIH-010-2	Sputum	2	S	2	S
NIH-010-3	Sputum	2	S	2	S
NIH-010-4	BAL	2	S	2	S
NIH-010-5	Sputum	4	S	2	S
NIH-010-6	Skinlesion(mouth)	<=1	S	1	S
NIH-011-1	Sputum	<=1	S	1	S
NIH-013-1	BAL	<=1	S	2	S
NIH-014-1	AXL	<=1	S	2	S
NIH-015-1	Blood	<=1	S	1	S
NIH-016-1	Sputum	>8	R	>16	R
NIH-016-2	Sputum	4	S	4	I^
NIH-016-3	Sputum	>8	R	16	R
NIH-017-1	Nephrostomy urine	<=1	S	1	S
NIH-018-1	Sputum	<=1	S	2	S
NIH-018-2	Sputum	2	S	2	S
NIH-018-3	BAL	2	S	2	S
NIH-019-1	Sinus	2	S	2	S
NIH-021-1	wound	2	S	1	S
NIH-022-1	Sputum	4	S	4	I^
NIH-023-1	BAL	2	S	2	S
NIH-025-1	Blood	>8	R	8	R
NIH-025-2	Blood	>8	R	8	R
NIH-025-3	Blood	>8	R	8	R
NIH-025-4	axl	>8	R	8	R
NIH-025-5	Blood	>8	R	4	I^
NIH-025-6	BAL	>8	R	8	R
NIH-025-7	Blood	>8	R	4	I^
NIH-026-1	Wound (l heel)	4	S	2	S

5. Discussion

A. xylosoxidans is an emerging pathogen that has garnered increasing attention in recent years. Accurate identification of *Achromobacter* species has been challenging due to limited knowledge, reliance on traditional and labor-intensive phenotypic identification methods, and the lack of comprehensive spectra from representative species within MALDI-TOF databases (74). Advances in sequencing technology and the decreasing cost of sequencing have made whole genome sequencing a feasible approach for identifying *A. xylosoxidans*. Mining these whole genome sequencing data can provide insights into the genetic diversity, antibiotic resistance mechanisms, and virulence potential of these isolates. In this study, we analyzed publicly available sequences from NCBI and our data from isolates collected at the NIH clinical centers representing various genetic backgrounds.

Our core genome analysis revealed that *A. xylosoxidans* possesses a relatively small core genome (less than 60% of its total genes). This is smaller than the core genomes of other opportunistic pathogens such as *P. aeruginosa* and *Helicobacter pylori*. For instance, *P. aeruginosa* has approximately 5,300 core genes (75) out of a total of ~ 5,700 genes (76), accounting for more than 92% of its genome. Similarly, *Helicobacter pylori* has 1,227 core genes (77) out of ~1,500 total genes (78), making up 81% of its genome. The core genome phylogenetic tree of *A. xylosoxidans* appears to be non-clonal and isolates from the same geographic locations or sample sources did not cluster together. The lack of population structure and co-ancestry suggests that *A. xylosoxidans* isolates are likely evolving independently, possibly under selective pressures in different environments or hosts. Such variations may arise from diverse soil or water environments (where *Achromobacter* are usually found), or from differences in antibiotics treatment regimens, underlying host diseases, or other health conditions. It is also worth noting that we do observe an exception of a few environmental isolates that do show some level of shared ancestry. Most isolates collected from the same patients clustered together and shared co-ancestry (Figures 2 and S1) suggesting that they evolved from the same infection. However, an exception was observed with NIH-005-4, which did not cluster with NIH-005-3 and NIH-005-2, indicating that NIH-005-2 and NIH-005-3 likely represent a case of reinfection rather than a persistent infection.

The multidrug resistance of *A. xylosoxidans* is common and concerning (Figure 3). This bacterium encodes several efflux pumps, although a few have been validated. The resistance-nodulation-cell division (RND)-type multidrug efflux pump, AxyABM, gave some level of resistant to cephalosporins (except cefepime), aztreonam, nalidicic acid, fluoroquinolones and chloramphenicol (79). Recently, a novel RND-type multidrug efflux pump, AxyXY-OprZ, was identified, which expels antibiotics such as cefepime, fluoroquinolones, tetracycline or

aminoglycosides including kanamycin, gentamicin, and tobramycin (80). Additionally, a recent study discovered the presence of another narrow-spectrum RND-type multidrug efflux pump, AxyEF-OprN, which pumps out fluoroquinolones, tetracyclines, and carbapenems (31). Furthermore, a very recent study identified AxySUV in some of the *A. xylosoxidans* genomes that export fluoroquinolones (ciprofloxacin, levofloxacin), carbapenems (doripenem), tetracycline (doxycycline, minocycline), and chloramphenicol (81). In addition to the presence of various efflux pumps, intrahost evolution may contribute to increasing antibiotic resistance over time. One interesting case is the NIH-016 series. NIH-016-3, collected seven months after NIH-016-2 from the same patient, exhibits higher AR and enhanced biofilm formation compared to NIH-016-2. Phylogenetic analysis confirms their close relationship, suggesting that NIH-016-3 evolved from NIH-016-2. While both isolates carry similar sets of resistance genes identified from the CARD database, NIH-016-3 demonstrates greater resistance to antibiotics. Comparing single nucleotide polymorphism of all three isolates in the series suggest that the later collected isolates accumulated mutations in the efflux pump (AxyXY-OprZ) and DNA mismatch repair systems that could facilitate the adaptation of the isolates to antibiotic treatment. Further study is needed to validate the roles of these SNPs.

In our NIH isolates, 21% of isolates are resistant to IMI while 34% isolates are resistant to IMR. Our models predict 20% resistant to IMR among NCBI isolates. Relebactam, a wide spectrum β − lactamase inhibitor against class A (extended spectrum β−lactamases KPC) and class C (AmpC) β−lactamases, does not seem to work in *Achromobacter*. In fact, all NIH isolates that were resistant to IMI were also resistant to IMR (Table S3). In other bacteria such as *Pseudomonas* and *Klebsiella*, IMR was more effective than IMI (82, 83). This could be because *A. xylosoxidans* possesses other β−lactamases classes. NIH-016s possess AXC family carbapenem-hydrolyzing class A β−lactamase and OXA-114 family class D β−lactamase. The OXA-114 may help *Achromobacter* to escape relebactam. These findings highlight that *A. xylosoxidans* employ multiple AR genes and strategies, which require further research .

As secretion systems are key virulent factors in bacteria, we investigated the presence of all possible secretion systems in the NIH isolates. T1SS, T2SS, T5SS, and bacterial Tight adherence Secretion System (TadSS) were identified in all isolates. T1SS is capable of transporting unfolded substrates, including digestive enzymes (proteases and lipases), adhesins, heme-binding proteins, and proteins with repeats-in-toxins (RTX) motifs (84). T2SS, a secretion machinery prevalent in many Gram-negative bacteria, facilitates the secretion of virulence factors such as proteases, toxins, lipases, and carbohydrate-degrading enzymes, which aid the survival of both pathogenic and environmental bacteria (85). T5SS is involved in the secretion of virulence proteins critical to pathogenesis in several bacteria, including immunoglobulin A protease of *Neisseria gonorrhoeae (86)*, IcsA protein of *Shigella flexneri* (87), and YadA of *Yersinia enterocolitica (88)*. TadSS is an important secretion system that supports infection, cell adherence, competence, and biofilm formation in many pathogenic bacteria, such as *P.*

aeruginosa, Bordetella pertussis, Burkholderia pseudomallei, and *Mycobacterium tuberculosis*
(89). The presence of these secretion systems in all isolates suggests that they all have the
potential to be pathogenic.

Most isolates have T3SS, T4SS, T6SS and flagella. T3SS is associated with the cytotoxicity of
the isolates (15). AxoU is the only known T3SS substrate that is responsible for host cell
cytotoxicity in *A. xylosoxidans*. All cytotoxic isolates carried both T3SS and AxoU, except NIH-
019-1 carries T3SS but not AxoU, suggesting the presence of unidentified T3SS substrate(s).
T4SS is primarily involved in horizontal genetic transfer (HGT) between different
microorganisms, which may contribute to pathogenesis and antibiotic resistance (16). T6SS is
a multiprotein complex that uses a spring-like mechanism to deliver effector proteins into
target cells (90). In *A. xylosoxidans,* T6SS plays a role in targeting competing bacteria and is
crucial for the internalization of the pathogen into lung epithelial cells (16). Flagella, which are
evolutionarily related to T3SS, play roles in motility, surface adherence and biofilm formation
(91). Interestingly, while most isolates encode flagellar genes, only 59% of them are motile,
suggesting an adaptation strategy of the chronic infection isolates. A study on the genotypic
and phenotypic evolution of the *A.xylosoxidans* isolates collected from CF patients over time
showed a decrease in swimming motility, an increase in antibiotic resistance, and enhanced
biofilm formation in evolved strains (92).

Phylogenetic analysis indicates that the T3SS of *A. xylosoxidans* structurally belongs to the Ysc
family of T3SS systems, which includes *Bordetella* spp. (alpha-proteobacteria), *Yersinia* spp.,
P. aeruginosa, *Aeromonas* spp., *Photorhabdus luminescens*, and *Vibrio* spp. (gamma-
proteobacteria) (Figure 4). In *Bordetella spp., Yersinia* spp., and *Photorhabdus luminescens,*
the T3SS confers resistance to the innate immune response by inhibiting phagocytosis, leading
to extracellular localization of the pathogens (93-95). While *P. aeruginosa* is known as an
extracellular pathogen, it can replicate intracellularly where T3SS and its effectors (mainly
ExoS) play critical roles in internalization, vacuole escape, evasion of inflammasome and
autophagy as well as bleb formation prior to detachment (96). Similarly, T3SS is important for
Aeromonas and *Vibrio* to maintain its intracellular niches (97, 98). In *A. xylosoxidans,* survival
within macrophages for several hours is independent of T3SS; however, cytotoxicity depends
on the presence of T3SS (12, 13). AxoU is the only known T3SS substrate linked to host cell
lysis (15), although other T3SS effectors may be secreted during infection, as disrupting the
T3SS (e.g., *sctV*) prevents cell death, whereas disruption of *axoU* alone does not (12).
Interestingly, T3SS and *axoU* are expressed shortly after infection (within 30 minutes), but are
subsequently turned off (12). Whether T3SS is reactivated when the bacteria escape the
vacuole and lyse the host cells remain to be tested. Whether there exist additional T3SS
effectors and what their roles are require further study.

Cytotoxicity demonstrates substantial heterogeneity among our clinical isolates (Figure 5). *A.*
xylosoxidans cytotoxicity is conducted through its T3SS with the major effector known to

cause cell death being AxoU, a homologue of ExoU found in *P. aeruginosa (99).* Interestingly,
our results would imply the presence of at least one additional effector which contributes to
cytotoxicity. While lacking *axoU* on its chromosome, NIH-019-1 demonstrated significantly
increased cytotoxicity compared to the GN008 negative control. The expression of T3SS only
occurs following phagocytosis and is not contact-dependent (13). There was no observable
relationship between overall flagellar motility and cytotoxicity but a small association with
biofilm production and cytotoxicity (Figure S2). In NIH-018 series, there was an increase in
cytotoxicity from isolates collected after the others (Table 2). This is interesting, as in other
bacteria, the bacteria tend to get less virulent when they colonized the host for a long time
(100, 101) .

Biofilm production is another major aspect of the bacterial lifestyle. Production of biofilm is
important in promoting bacterial resistance to environmental pressures and defending against
immune cells during infection. The NIH isolates produce low level of biofilm in general (Figure
6). We observed a slight increase in biofilm production in the intrahost isolates, NIH-010s, NIH-
016s and NIH-018s but it is not significant. GN050 collected from ear effusion (15), is the only
isolate that produced robust biofilm. It could be that because most of our isolates were from
respiratory tract, they don't form as much biofilm. Alternatively, biofilm was tightly regulated,
and we didn't assay the bacteria in the desire condition for biofilm formation in these
particular isolates. Biofilm production was not correlated with motility (Figure S2), although
the strong biofilm former are moderately motile. Biofilm was not correlated with antibiotic
resistance, indicating these phenotypes operate through independent mechanisms.

Flagellar motility in bacteria is commonly associated with increased disease severity and
pathogenesis. Multiple related species such as *Bordetella* and *Pseudomonas* are known to
have the expression of virulence systems such as the T3SS modulated by the expression of the
flagella (65, 102) . Additionally, bacteria with dysfunctional flagella are known to have reduced
pathogenicity *in vivo* due to the inability to reach the site of infection (64). Bacteria isolated
from the lungs are known to preferentially gain mutations in the flagella resulting in non-
motile genotypes as compared to other sites of infection (60) (66). Our results demonstrated
this trend. We identified a mutation in *fliF* which we could result in a loss of flagellar
production. We observed no correlation between overall motility and cytotoxicity (Figure S2).
Interestingly, isolates that are the most cytotoxic are motile, but the most motile isolates are
not cytotoxic. It could be possible that the most motile isolates can escape phagocytosis in
our *invitro* infection. Strains that experience a loss of either flagellar motility or cytotoxicity
do not experience a relative loss in the other. It is unclear what role the flagella of *A.*
xylosoxidans plays during infection. In multiple bacteria species the function of the flagella is
highly correlated with the activity of virulence mechanisms. The regulatory relationship
between motility and virulence varies substantially between organisms. *Bordetella spp.*

display an inverse relationship (102), whereas *Pseudomonas aeruginosa* displays a positive relationship between T3SS production and flagellar motility (65). *Salmonella spp.* also demonstrate a positive relationship with the SPI-1 T3SS but not the SPI-2 T3SS (103). These observations indicate a general positive correlation between flagella and virulence factor expression in extracellular pathogens. In *A. xylosoxidans,* T3SS expression seemed to increase shortly after the bacteria are phagocytosed by macrophage but dampened quickly. Whether T3SS is shut off and then gets re-activated at later stage of infection or it remains functional to some extend without making more apparatus and elevated level of substrate production requires further investigations.

In summary, this study highlights the genetic and phenotypic diversity among *A. xylosoxidans* isolates, as evidenced by the high accessory genome content and variability in secretion systems. Significant differences in antibiotic resistance and virulence mechanisms underscore the adaptability of this bacterial species. These findings emphasize the need for further mechanistic studies to elucidate the pathogenicity and antibiotic resistance mechanisms of *A. xylosoxidans*.

6.   Author statements

Pooja Acharya (Formal Analysis, Data Curation, Methodology, Writing), Cameron Lloyd (Formal Analysis, Methodology, Writing), Ngoc Thien Lam (Formal Analysis, Writing), Jessica Kumke (Formal Analysis), Sreejana Ray (Resources), Zilia Yanira Muñoz Ramirez (Formal Analysis), Sanjay Mandal (Formal analysis), Sanchita Das (conceptualization, resources) and Hanh Ngoc Lam (Conceptualization, Data Curation, Formal Analysis, Funding Acquisition, Investigation, Methodology, Supervision, Writing)

6.1   Conflicts of interest

The author(s) declare that there are no conflicts of interest

6.2   Funding information

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6.3   Acknowledgements

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