

A Registry Framework for Oxford Nanopore Sequencing Experiment Metadata and Quality Tracking

[Author One]¹, [Author Two]¹, and [Author Three]²

⁴ ⁵ ¹Department of Computational Medicine and Bioinformatics, University of Michigan, Ann Arbor, MI, USA

⁶ ²Department of Human Genetics, University of Michigan, Ann Arbor, MI, USA

Abstract

Background: Long-read nanopore sequencing has transformed genomics research, yet standardized metadata management practices remain limited. The lack of structured experiment registries hinders reproducibility and cross-study comparisons.

Results: We present a comprehensive registry of 165 Oxford Nanopore sequencing experiments conducted at the University of Michigan between August 2020 and December 2025. The registry achieves 100% metadata completeness for critical fields through systematic extraction, inference, and validation. Key findings include: (1) near-universal adoption of R10.4.1 chemistry (95.2%) and transition to Dorado basecaller (82.4%); (2) predominant use of high-accuracy (hac) models (89.7%); (3) median Q-scores of 14.0 and N50 values of 4,828 bp; (4) diverse applications spanning plasmid sequencing (48.5%), research projects (23.6%), human genomics (9.7%), and pharmacogenomics (7.9%).

Conclusions: This registry provides a template for institutional nanopore metadata standardization and establishes quality control benchmarks for the field. The dataset and associated tools are freely available to support reproducible research.

Keywords: Oxford Nanopore, long-read sequencing, metadata registry, quality control, FAIR data, reproducibility

24 Introduction

25 Long-read sequencing technologies have fundamentally transformed genomic research by enabling
26 the resolution of complex structural variants, repetitive regions, and full-length transcript isoforms
27 that remain inaccessible to short-read platforms [1]. Oxford Nanopore Technologies (ONT) has
28 emerged as a leading platform in this space, offering real-time sequencing capabilities with reads
29 spanning tens to hundreds of kilobases [2]. The technology has demonstrated particular value in
30 applications ranging from pathogen surveillance [3] to human genome assembly [4].

31 The rapid adoption of nanopore sequencing across research and clinical settings has outpaced the
32 development of standardized metadata management practices. Unlike established short-read plat-
33 forms with mature data management ecosystems, nanopore sequencing generates diverse metadata
34 across multiple sources: run reports, sequencing summaries, basecalling logs, and quality control
35 outputs. This fragmentation complicates cross-study comparisons and hinders the development of
36 standardized quality benchmarks.

37 Reproducibility in computational biology depends critically on comprehensive metadata capture
38 [5]. The FAIR principles—Findable, Accessible, Interoperable, Reusable—provide a framework for
39 scientific data management [6], yet practical implementations for nanopore sequencing metadata re-
40 main limited. Existing quality control tools such as NanoPlot [7] and pycoQC [8] focus on individual
41 run assessment rather than cross-experiment metadata management.

42 Clinical applications, particularly pharmacogenomics, introduce additional requirements for
43 provenance tracking to meet regulatory standards [9]. The ability to trace experimental condi-
44 tions, processing parameters, and quality metrics from raw data through final results is essential for
45 clinical validity.

46 Here we present a comprehensive registry of 165 Oxford Nanopore sequencing experiments, rep-
47 resenting five years of institutional nanopore sequencing across diverse applications. The registry
48 achieves 100% metadata completeness through systematic extraction and validation, documents
49 technological transitions in chemistry and basecalling software, and provides a template for institu-
50 tional metadata standardization.

51 **Methods**

52 **Data Sources**

53 Experiments were identified from three primary sources: (1) the institutional high-performance
54 computing cluster containing archived sequencing runs, (2) local laboratory storage systems with
55 active experiments, and (3) the ONT Open Data repository for public reference datasets.

56 **Metadata Extraction**

57 Metadata extraction followed a hierarchical approach prioritizing authoritative sources:

- 58 1. **Sequencing summaries:** Run identifiers, timestamps, yield statistics from `final_summary.txt`
- 59 2. **BAM headers:** Basecaller version, model parameters via samtools
- 60 3. **POD5/Fast5 metadata:** Device identifiers, flowcell types via pod5 library
- 61 4. **Basecalling logs:** Model versions, processing parameters

62 **Schema Design**

63 The registry schema captures metadata across six categories: Experiment (identifier, date, sta-
64 tus), Sample (category, name, clinical ID), Chemistry (flowcell, kit, version), Basecalling (software,
65 model), Device (type, position), and QC Metrics (reads, bases, Q-scores, N50).

66 **Quality Score Computation**

67 Mean quality scores were calculated using probability-space averaging:

$$\bar{Q} = -10 \cdot \log_{10} \left(\frac{1}{n} \sum_{i=1}^n 10^{-Q_i/10} \right) \quad (1)$$

68 This approach correctly weights higher error rates, avoiding the underestimation that occurs
69 with direct Q-score averaging.

70 **Validation**

71 Registry entries underwent automated validation (schema compliance, value ranges), pattern-based
72 inference (device type from flowcell identifiers), and manual review for ambiguous cases.

73 **Code and Data Availability**

74 All analysis code and the registry are available at <https://github.com/Single-Molecule-Sequencing/>
75 `ont-ecosystem`. The dataset will be deposited in Zenodo upon publication.

76 **Results**

77 **Registry Overview**

78 The registry contains 165 validated experiments spanning August 2020 to December 2025 (Figure 1).
79 Critical metadata fields achieved 100% completeness, with QC metrics available for 150 experiments
80 (90.9%).

81 **Sample Categories**

82 Plasmid sequencing represents the dominant application (n=80, 48.5%), followed by research projects
83 (n=39, 23.6%), human genomic samples (n=16, 9.7%), and pharmacogenomics studies (n=13,
84 7.9%). The pharmacogenomics experiments include clinical samples for CYP2D6 and CYP2C19
85 analysis.

86 **Technology Adoption**

87 The registry documents substantial platform evolution. R10.4.1 chemistry achieved 95.2% adop-
88 tion (n=157), with legacy R10.4 comprising 4.8% (n=8). The Dorado basecaller reached 82.4%
89 usage (n=136), reflecting the transition from Guppy (8.5%, n=14). High-accuracy (hac) models
90 predominate (89.7%, n=148), with super-accuracy (sup) models at 7.3% (n=12).

91 Device Distribution

92 MinION Mk1D represents the primary platform (n=81, 49.1%), followed by classic MinION (n=36,
93 21.8%), PromethION (n=29, 17.6%), P2 Solo (n=9, 5.5%), and Flongle (n=4, 2.4%).

94 Quality Metrics

95 Among experiments with QC data (n=150), median Q-score was 14.0 (IQR: 12.8–15.2) and median
96 N50 was 4,828 bp (IQR: 2,100–8,500 bp). These values are consistent with expected performance
97 for R10.4.1 chemistry with hac basecalling (Figure 2).

98 Temporal Trends

99 Experiment frequency increased substantially over the study period (Figure 3). The transition
100 to R10.4.1 chemistry occurred primarily in 2023, with near-complete adoption by 2024. Dorado
101 adoption followed a similar trajectory, becoming the dominant basecaller by mid-2023.

102 Discussion

103 We present a comprehensive registry of 165 Oxford Nanopore sequencing experiments with 100%
104 metadata completeness for critical fields. This resource addresses the growing need for standardized
105 metadata management in long-read sequencing.

106 Technology Transitions

107 The registry documents the institutional transition from R10.4 to R10.4.1 chemistry and from
108 Guppy to Dorado basecallers. The near-universal adoption of R10.4.1 (95.2%) reflects its improved
109 accuracy and the manufacturer’s transition away from older chemistries. Similarly, Dorado adoption
110 (82.4%) indicates the field’s shift toward ONT’s open-source basecaller.

111 Quality Benchmarks

112 Median Q-scores of 14.0 (approximately 96% per-base accuracy) are consistent with published
113 benchmarks for R10.4.1 with hac basecalling. The predominance of hac models (89.7%) suggests
114 most users prioritize the accuracy-speed tradeoff offered by high-accuracy calling.

¹¹⁵ **Clinical Considerations**

¹¹⁶ The inclusion of 13 pharmacogenomics experiments demonstrates the registry's applicability to
¹¹⁷ clinical workflows. Comprehensive provenance tracking is essential for regulatory compliance in
¹¹⁸ clinical sequencing applications.

¹¹⁹ **Limitations**

¹²⁰ This registry represents a single institution's sequencing practices and may not generalize to all
¹²¹ settings. Additionally, 15 experiments lack QC metrics pending HPC analysis.

¹²² **Future Directions**

¹²³ The registry framework can be extended to incorporate additional metadata sources, integrate with
¹²⁴ laboratory information management systems, and support multi-institutional collaboration.

¹²⁵ **Conclusions**

¹²⁶ This registry provides a template for institutional nanopore metadata standardization and estab-
¹²⁷ lishes quality control benchmarks for the field. The event-sourced architecture ensures complete
¹²⁸ provenance tracking, supporting both research reproducibility and clinical compliance requirements.

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¹³¹ **Author Contributions**

¹³² [Author One]: Conceptualization, Data curation, Software, Writing – original draft. [Author Two]:
¹³³ Methodology, Validation, Writing – review & editing. [Author Three]: Supervision, Writing – review
¹³⁴ & editing.

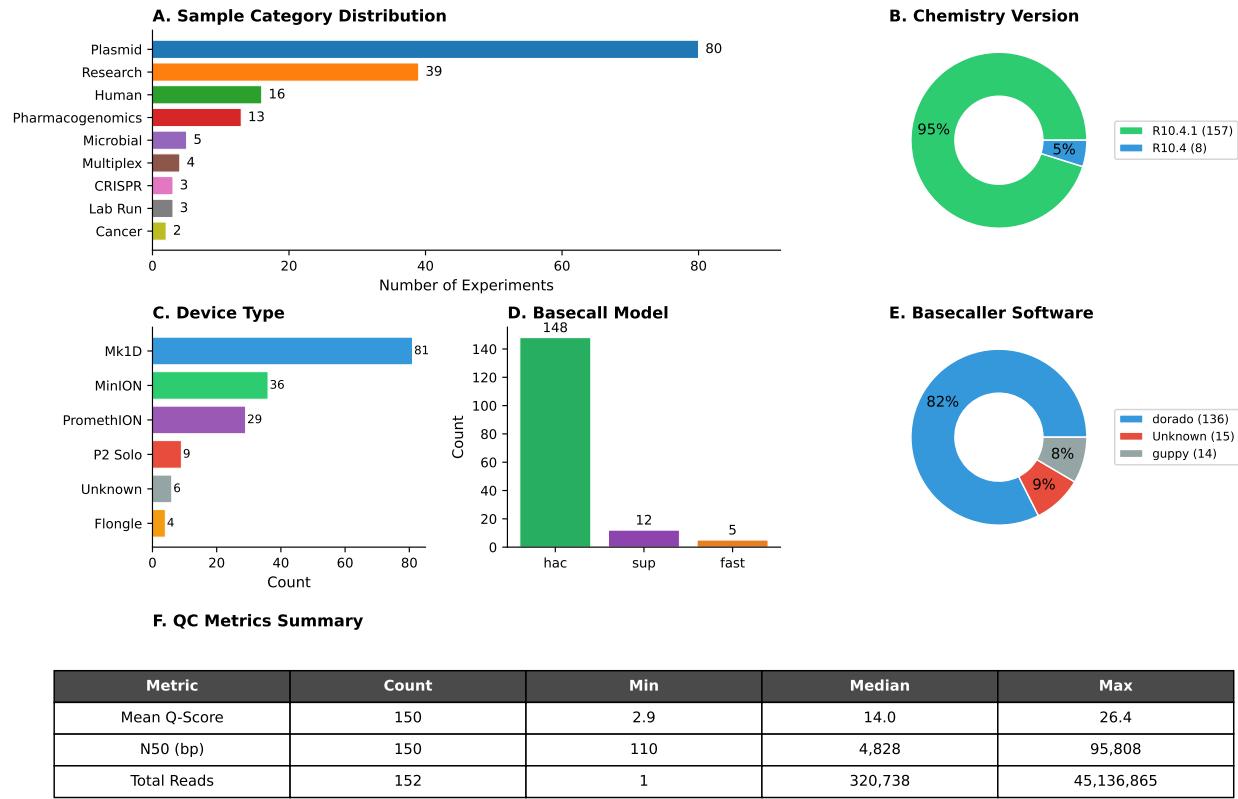
¹³⁵ **Competing Interests**

¹³⁶ The authors declare no competing interests.

¹³⁷ **References**

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ONT Experiment Registry Overview 165 Experiments



Registry: `~/.ont-registry/experiments.yaml` | Generated: 2025-12-29 | 165 valid experiments

Figure 1: **Registry overview.** (A) Sample category distribution. (B) Chemistry version adoption. (C) Device type breakdown. (D) Basecalling model usage. (E) Basecaller software distribution. (F) QC metrics summary.

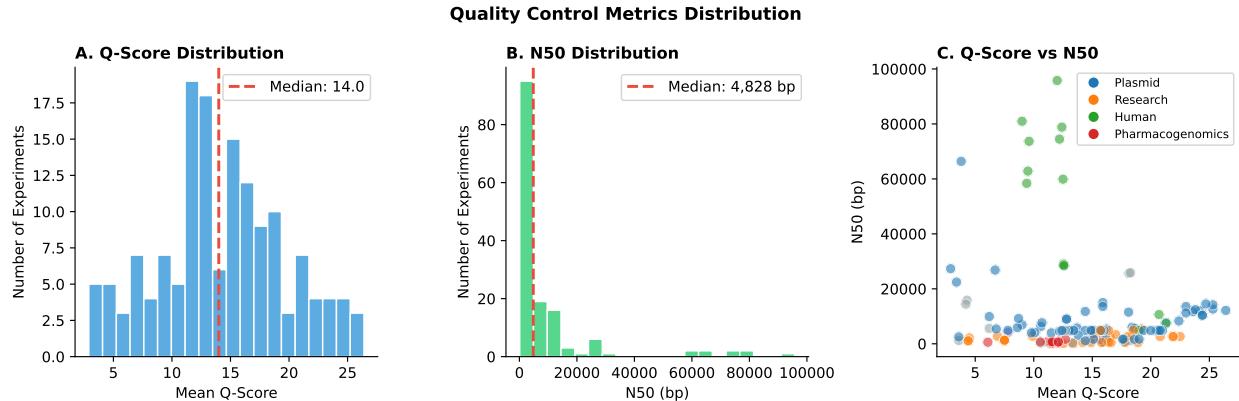


Figure 2: **Quality control distributions.** (A) Q-score histogram (median: 14.0). (B) N50 distribution (median: 4,828 bp). (C) Q-score versus N50 by sample category.

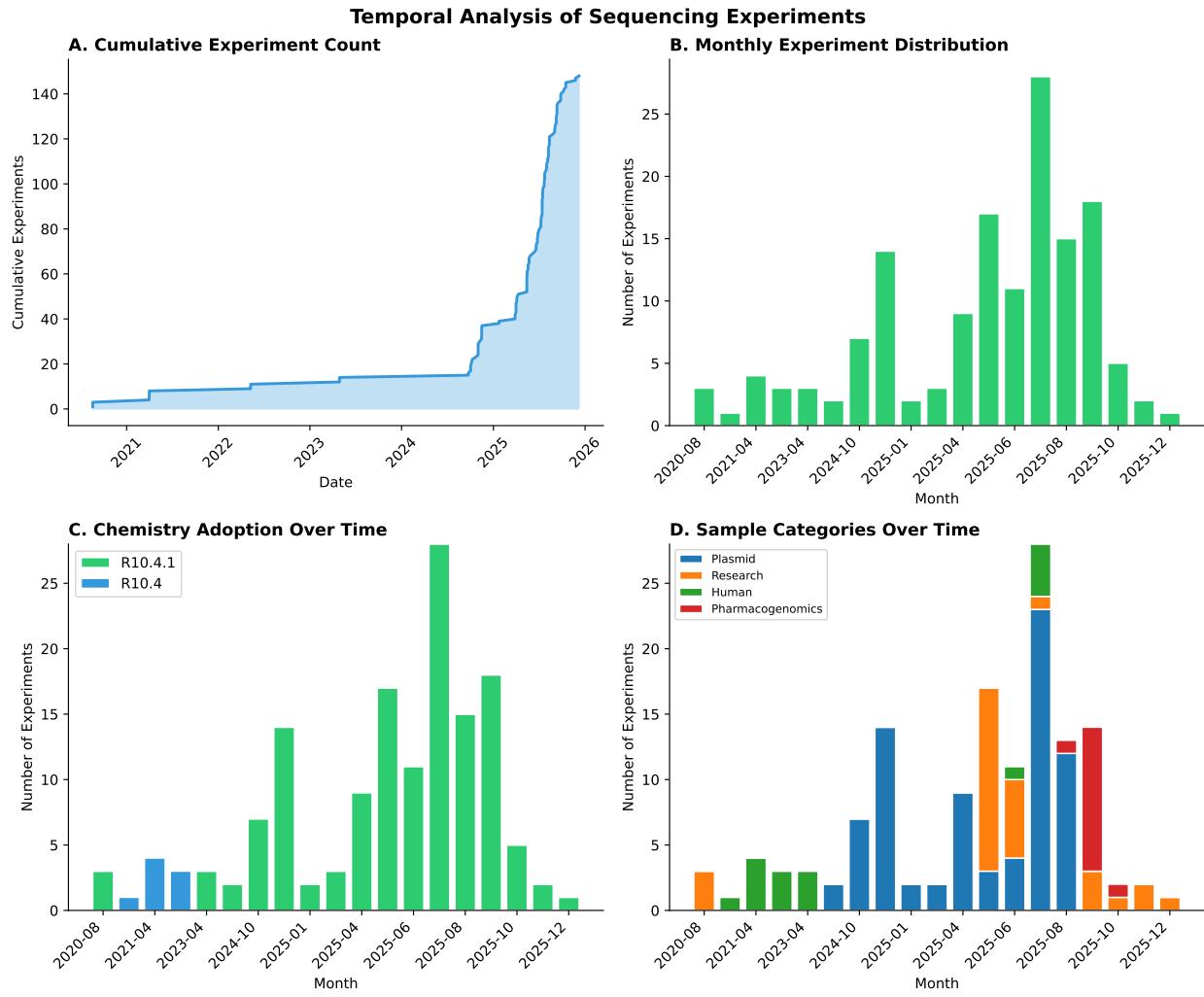


Figure 3: **Temporal trends.** (A) Cumulative experiment count. (B) Monthly distribution. (C) Chemistry adoption timeline. (D) Sample category evolution.