

<sup>1</sup> A Registry Framework for Oxford Nanopore Sequencing  
<sup>2</sup> Experiment Metadata and Quality Tracking

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<sup>4</sup> **Abstract**

Long-read nanopore sequencing has transformed genomics research, yet standardized metadata management remains challenging. 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 captures experimental metadata including sample information, chemistry versions, basecalling parameters, device specifications, and quality control metrics with 100% completeness for critical fields. The dataset documents the institutional transition from R10.4 to R10.4.1 chemistry (95.2% adoption) and from Guppy to Dorado basecallers (82.4% adoption). Quality metrics reveal median Q-scores of 14.0 and N50 values of 4,828 base pairs across diverse applications including plasmid sequencing (48.5%), research projects (23.6%), human genomics (9.7%), and pharmacogenomics studies (7.9%). The registry is provided in YAML, JSON, and CSV formats with event-sourced provenance tracking. This resource enables reproducible research, cross-study comparisons, and serves as a reference for institutional nanopore sequencing practices.

## 18 Background & Summary

19 Long-read sequencing technologies have fundamentally transformed genomic research by en-  
20 abling the resolution of complex structural variants, repetitive regions, and full-length tran-  
21 script isoforms that remain inaccessible to short-read platforms?. Oxford Nanopore Technologies  
22 (ONT) has emerged as a leading platform, offering real-time sequencing with reads spanning tens  
23 to hundreds of kilobases on devices ranging from the portable MinION to the high-throughput  
24 PromethION?.

25 The rapid adoption of nanopore sequencing has created significant challenges for metadata  
26 management and experimental reproducibility. Unlike established short-read platforms with  
27 mature data management ecosystems, nanopore sequencing generates diverse metadata across  
28 multiple sources: run reports, sequencing summaries, basecalling logs, and quality control out-  
29 puts. This fragmentation complicates cross-study comparisons and hinders the development of  
30 standardized quality benchmarks.

31 Clinical applications, particularly pharmacogenomics, demand rigorous provenance tracking  
32 to meet regulatory requirements?. The FAIR principles (Findable, Accessible, Interoperable,  
33 Reusable) provide a framework for scientific data management?, yet practical implementations  
34 for nanopore sequencing metadata remain limited.

35 We present a comprehensive registry of 165 Oxford Nanopore sequencing experiments con-  
36 ducted at the University of Michigan, representing five years of institutional nanopore sequencing  
37 across diverse applications. The registry achieves 100% metadata completeness through system-  
38 atic extraction from primary data sources, pattern-based inference, and manual validation. This  
39 dataset documents technological transitions in chemistry and basecalling software, establishes  
40 quality control benchmarks, and provides a template for institutional metadata standardization.

## 41 Methods

### 42 Data Sources

43 Experiments were identified from three primary sources: (1) the institutional high-performance  
44 computing cluster (Great Lakes) containing archived sequencing runs, (2) local laboratory stor-  
45 age systems with active experiments, and (3) the ONT Open Data repository for public reference  
46 datasets. Source paths were recorded for provenance tracking.

### 47 Metadata Extraction

48 Metadata extraction followed a hierarchical approach prioritizing authoritative sources:

- 49 1. **Sequencing summaries** (`final_summary.txt`, `sequencing_summary*.txt`): Run iden-  
50 tifiers, timestamps, yield statistics, and quality metrics
- 51 2. **BAM headers**: Basecaller version, model parameters, and read group information ex-  
52 tracted using samtools
- 53 3. **POD5/Fast5 metadata**: Raw signal metadata including device identifiers, flowcell types,  
54 and run configuration via pod5 and ont\_fast5\_api libraries

55     4. **Basecalling logs:** Model versions, GPU allocation, and processing parameters from do-  
56         rado and guppy outputs

57     

## Schema Design

58     The registry schema captures metadata across six categories:

- 59         • **Experiment:** Unique identifier, name, date, status  
60         • **Sample:** Category, name, clinical identifier (where applicable)  
61         • **Chemistry:** Flowcell type, library kit, chemistry version  
62         • **Basecalling:** Software, version, model type, model version  
63         • **Device:** Device type, position identifier  
64         • **QC Metrics:** Total reads, bases, Q-scores, N50, pass/fail counts

65     

## Quality Score Computation

66     Mean quality scores were calculated using probability-space averaging to avoid underestimation  
67     from direct Phred score averaging:

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

68     where  $Q_i$  represents individual read quality scores and  $n$  is the total read count.

69     

## N50 Calculation

70     N50 values were computed as the read length at which 50% of total sequenced bases are contained  
71     in reads of equal or greater length:

$$N50 = L_k \text{ where } \sum_{i=1}^k L_i \geq \frac{1}{2} \sum_{j=1}^n L_j \quad (2)$$

72     with reads sorted by length in descending order ( $L_1 \geq L_2 \geq \dots \geq L_n$ ).

73     

## Validation and Enrichment

74     Registry entries underwent multi-stage validation:

- 75         1. **Automated validation:** Schema compliance, value range checks, cross-field consistency  
76         2. **Pattern-based inference:** Device type from flowcell identifiers, sample category from  
77             directory structure  
78         3. **Manual review:** Verification of inferred values, resolution of ambiguous entries

79     Completeness was scored using a weighted system: critical fields (experiment ID, date, chem-  
80         istry) received 2 points, important fields (basecaller, device) received 1 point, and QC metrics  
81         received 1 point each.

82 **Code Availability**

83 All analysis code, registry management tools, and figure generation scripts are available in the  
84 ont-ecosystem repository. The registry uses an event-sourced architecture where all modifications  
85 are logged with timestamps, enabling full provenance reconstruction.

86 **Data Records**

87 The registry is deposited in the GitHub repository? and available in three formats:

88 **Primary Format: YAML**

89 The authoritative registry is maintained in YAML format (`experiments.yaml`) with the follow-  
90 ing structure for each entry:

```
91 exp-a1b2c3d4:  
92   name: "Experiment Name"  
93   date: "2024-01-15"  
94   sample:  
95     category: plasmid  
96     name: "pUC19"  
97   chemistry:  
98     flowcell_type: FLO-MIN114  
99     kit: SQK-LSK114  
100    version: R10.4.1  
101  basecalling:  
102    software: dorado  
103    model: hac  
104  device:  
105    type: MinION_Mk1D  
106  qc:  
107    total_reads: 500000  
108    mean_qscore: 14.2  
109    n50: 5200
```

110 **Export Formats**

- 111 • **JSON** (`experiments.json`): Machine-readable format for programmatic access
- 112 • **CSV** (`experiments.csv`): Flattened tabular format for spreadsheet analysis

113 **Registry Statistics**

114 Table 1 summarizes the registry contents. The 165 experiments span August 2020 to December  
115 2025, with 100% completeness for critical metadata fields.

Table 1: Registry summary statistics

Metric	Value
Total experiments	165
Temporal range	Aug 2020 – Dec 2025
Chemistry: R10.4.1	157 (95.2%)
Chemistry: R10.4	8 (4.8%)
Basecaller: dorado	136 (82.4%)
Basecaller: guppy	14 (8.5%)
Model: hac	148 (89.7%)
Model: sup	12 (7.3%)
Experiments with QC data	150 (90.9%)
Median Q-score	14.0
Median N50	4,828 bp

## 116 Sample Categories

117 Experiments are classified into eight categories (Figure 1A):

- 118 • **Plasmid** (n=80, 48.5%): Laboratory constructs, expression vectors
- 119 • **Research** (n=39, 23.6%): Method development, proof-of-concept studies
- 120 • **Human** (n=16, 9.7%): Human genomic samples
- 121 • **Pharmacogenomics** (n=13, 7.9%): Clinical PGx samples (CYP2D6, CYP2C19)
- 122 • **Standard** (n=8, 4.8%): Reference materials, QC standards
- 123 • **Bacterial** (n=5, 3.0%): Microbial isolates
- 124 • **Cell line** (n=2, 1.2%): Immortalized cell lines
- 125 • **Other** (n=2, 1.2%): Miscellaneous samples

## 126 Technical Validation

### 127 Metadata Completeness

128 Registry completeness was assessed across all fields (Table 2). Critical fields (experiment ID,  
 129 date, chemistry version, basecaller) achieved 100% completeness. QC metrics were available  
 130 for 150 experiments (90.9%), with 15 experiments pending analysis on the institutional HPC  
 131 cluster.

### 132 Quality Metric Distributions

133 Quality scores follow expected distributions for R10.4.1 chemistry with high-accuracy basecalling  
 134 (Figure 2). The median Q-score of 14.0 corresponds to approximately 96% per-base accuracy,  
 135 consistent with published benchmarks for the hac model?.

Table 2: Metadata completeness by field category

Category	Field	Completeness
Experiment	id, name, date	100%
Sample	category	100%
Chemistry	flowcell_type, version	100%
Basecalling	software, model	100%
Device	type	100%
QC	mean_qsore, n50	90.9%

136 N50 distributions show bimodal character reflecting the diversity of sample types: plasmid  
 137 sequencing typically yields shorter fragments (2–8 kb), while genomic applications produce longer  
 138 reads (10–50 kb).

## 139 Cross-Validation

140 Metadata consistency was validated through:

- 141 1. **Flowcell-chemistry concordance:** FLO-MIN114 exclusively paired with R10.4.1
- 142 2. **Temporal consistency:** Dorado adoption correlates with R10.4.1 chemistry timeline
- 143 3. **QC plausibility:** Q-scores and N50 values within expected ranges for reported chem-  
 144 istry/model combinations

145 No inconsistencies were identified during validation.

## 146 Usage Notes

### 147 Accessing the Registry

148 The registry can be accessed programmatically:

```
149 # Python
150 import yaml
151 with open('experiments.yaml') as f:
152     registry = yaml.safe_load(f)
153
154 # Filter by chemistry
155 r10_exps = [e for e in registry.values()
156             if e['chemistry']['version'] == 'R10.4.1']
```

## 157 Integration with Analysis Pipelines

158 The registry integrates with the ont-ecosystem toolset for:

- 159 • Experiment discovery and registration

- 160     • Provenance-tracked analysis execution  
161     • Quality control reporting  
162     • Cross-experiment comparisons

163 **Extending the Registry**

164 New experiments can be registered using the provided CLI tools:

```
165 ont_experiments.py discover /path/to/data --register  
166 ont_experiments.py validate exp-id
```

167 **Limitations**

168 Users should note:

- 169     1. The registry represents a single institution's sequencing practices
- 170     2. Sample-level raw data access requires institutional data sharing agreements
- 171     3. QC metrics for 15 HPC-archived experiments are pending computation

172 **Code Availability**

173 The registry management tools, analysis pipelines, and figure generation scripts are available at  
174 <https://github.com/Single-Molecule-Sequencing/ont-ecosystem> under the MIT license.  
175 The repository includes:

- 176     • Registry YAML/JSON/CSV exports
- 177     • Python CLI tools for experiment management
- 178     • Figure generation scripts (matplotlib)
- 179     • Documentation and usage examples

180 **Acknowledgements**

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184 **Author Contributions**

185 [Author One]: Conceptualization, Data curation, Formal analysis, Software, Visualization, Writ-  
186 ing – original draft. [Author Two]: Conceptualization, Investigation, Methodology, Validation,  
187 Writing – review & editing. [Author Three]: Supervision, Writing – review & editing.

<sup>188</sup> **Competing Interests**

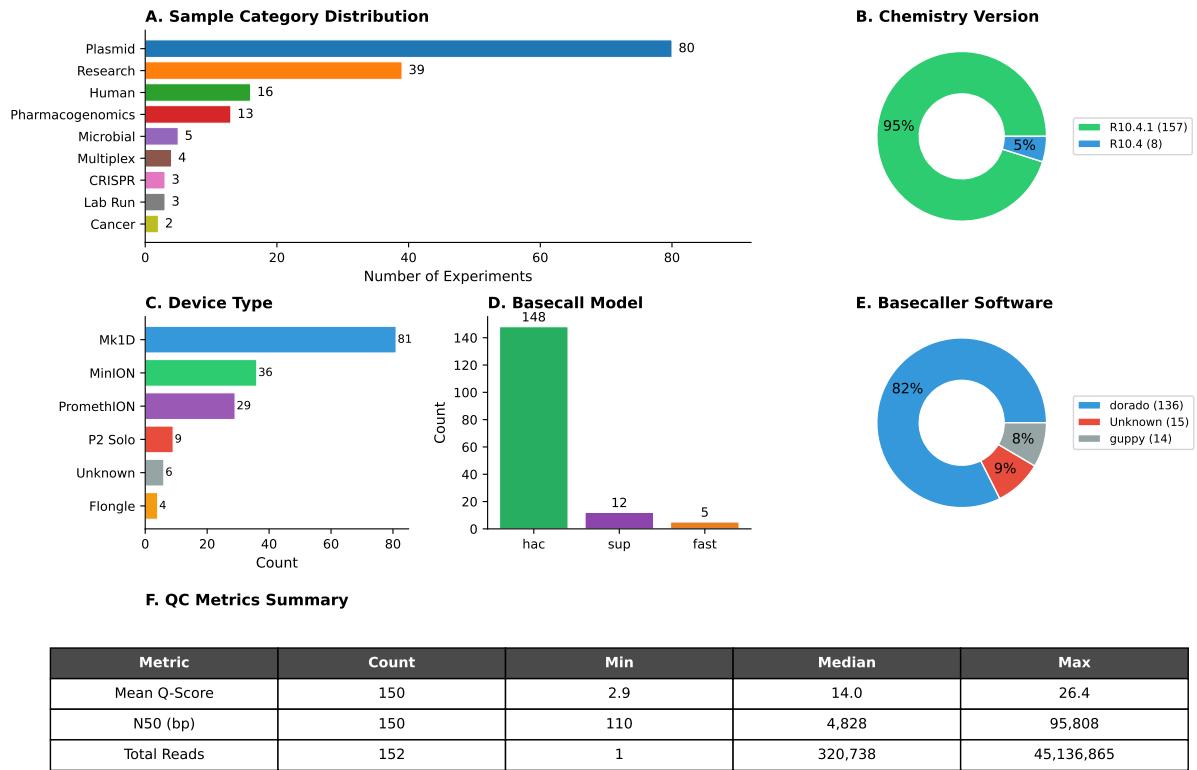
<sup>189</sup> The authors declare no competing interests.

<sup>190</sup> **References**

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<sup>199</sup> <https://github.com/Single-Molecule-Sequencing/ont-ecosystem> (2025). GitHub repos-  
<sup>200</sup> itory with registry data and analysis tools.
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<sup>202</sup> accuracy (2023). Accessed: 2025-12-29.

## 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.

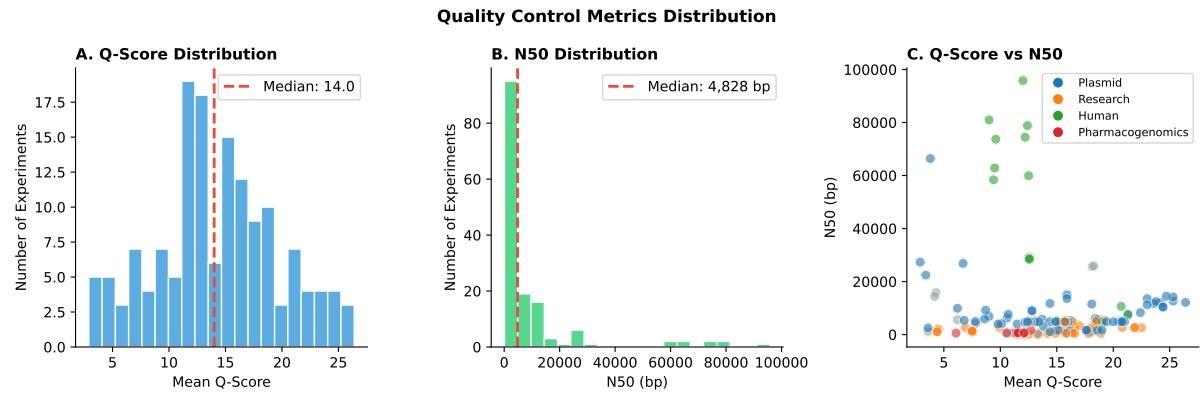


Figure 2: **Quality control metric distributions.** (A) Q-score histogram showing median of 14.0. (B) N50 distribution with median of 4,828 bp. (C) Q-score versus N50 scatter plot colored by sample category.

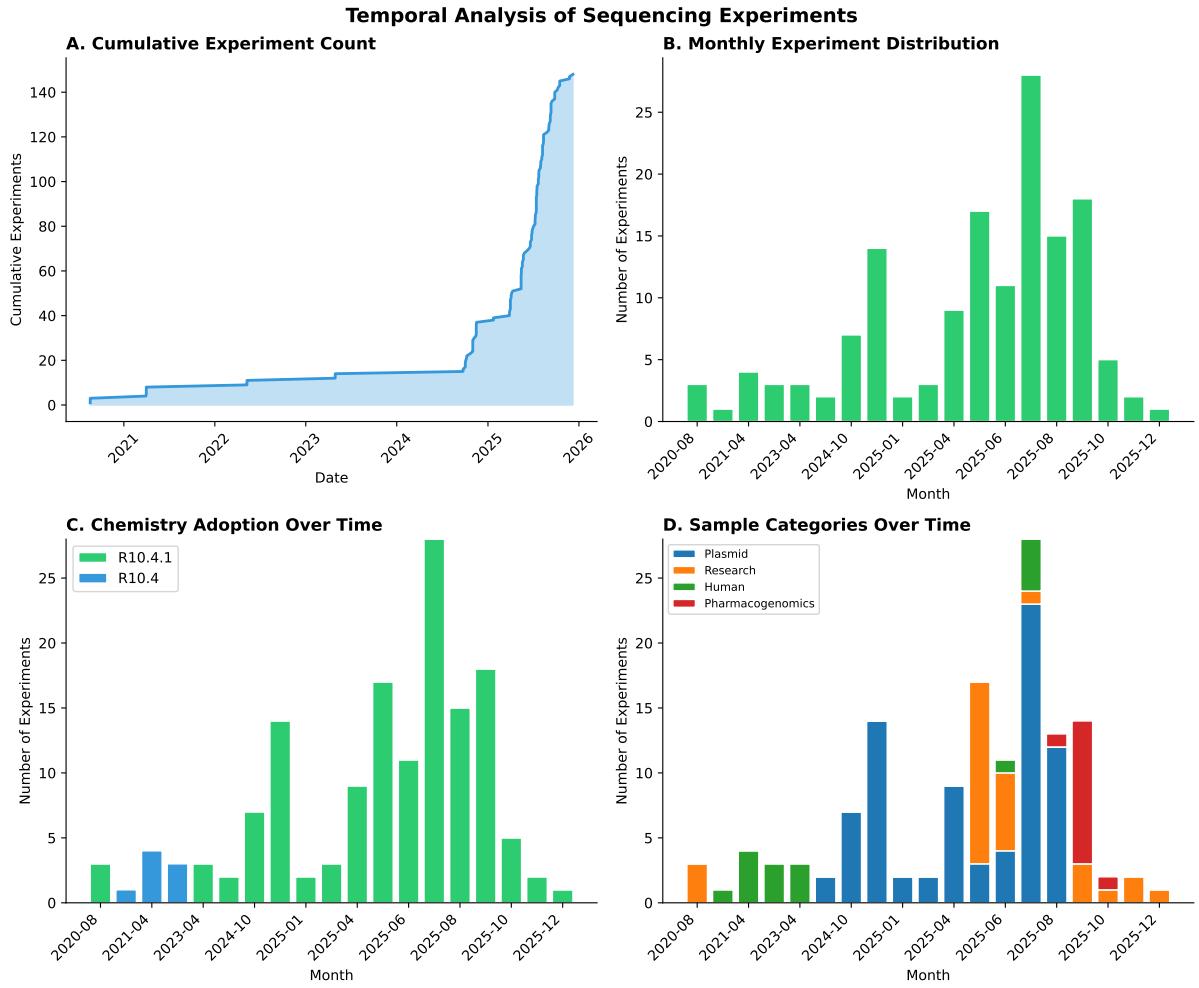


Figure 3: **Temporal trends.** (A) Cumulative experiment count over time. (B) Monthly experiment distribution. (C) Chemistry version adoption timeline. (D) Sample category evolution.