

DPath: Decentralized Pathology AI for Global Health Equity

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TL;DR

DPath is a decentralized AI platform tackling global cancer diagnostic disparities by leveraging **synthetic data and NFTs** to enable hospitals and pathology labs to securely share pathology data while retaining full ownership. Focusing first on prostate and endometrial cancers, it trains equitable AI models exclusively on **privacy-preserving synthetic biopsies**, reducing diagnostic errors by **≥30%** and turnaround times from weeks to days.

Contributors—hospitals/labs, data annotators, and model developers—earn royalties through smart contracts, while synthetic data generation, zero-knowledge proofs, and bias audits ensure ethical deployment. By 2030, DPath aims to empower **1,000+ hospitals/labs** with **10+ cancer diagnostics AI** globally.

1. Introduction

1.1 Privacy-Centric Diagnostic Gaps

Low- and Middle-Income Countries (LMICs) face two critical barriers:

1. **Data Privacy Risks:** 78% of LMIC hospitals avoid sharing patient data due to privacy concerns and regulatory gaps **[1]**.
2. **Diagnostic Delays:** Prostate cancer misdiagnosis rates exceed 40% in regions with pathologists per 100,000 people **[2]**.

1.2 Synthetic-First Architecture

DPath's **privacy-by-design** approach:

- Hospitals/labs **never share raw patient data**.
 - Locally trained diffusion models generate synthetic biopsies that preserve diagnostic features but eliminate patient-specific identifiers **[3]**.
 - Synthetic data is tokenized as NFTs for annotation, AI training, and global sharing.
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2. Core Components

2.1 Synthetic Data Workflow

- Local Synthetic Training:**
 - Hospitals use lightweight diffusion models (e.g., Stable Diffusion LoRA) trained on their raw data to generate synthetic H&E slides [4].
 - Validated to retain >95% diagnostic accuracy vs. real biopsies [5].
- Synthetic NFT Minting:**
 - Synthetic slides are tokenized on Ethereum, embedding metadata (e.g., cancer subtype, stain type).
- Downstream Use:**
 - Annotators label synthetic NFTs.
 - AI developers train models on synthetic datasets.

Case Example:

A Malaysian hospital generated 1,000 synthetic prostate biopsies from 50 real cases. The synthetic NFTs achieved 98% diagnostic concordance in external validation, with zero patient data exposure [6].

2.2 Contributor Roles & Incentives

Role	Actions	Ownership & Earnings
Hospitals/Labs	Generate synthetic NFTs from local data.	Own 100% of raw data + 50% NFT royalties.
Data Annotators	Label synthetic NFTs (e.g., Gleason grades).	Earn 10% NFT royalties + cash/token rewards.
Model Developers	Train AI on synthetic NFT datasets.	30% model ownership; 70% to NFT contributors.

3. Technology Stack

3.1 Privacy-Preserving Synthetic Pipeline

- **Diffusion Models:** Generate synthetic biopsies locally by training diffusion-based synthetic image models on-site.
- **NFT Provenance:** Zero-knowledge proofs (ZKPs) verify synthetic data lineage without exposing raw data sources [8].
- **Evaluation:** AI models are tested on local hospital data to ensure synthetic-to-real generalizability [9].

3.2 AI Marketplace

- **Synthetic Data Pools:** Global repository of prostate/endometrial synthetic NFTs, tagged by subtype and demographics.
 - **Royalty Automation:** Smart contracts split subscription fees:
 - 42% to contributors (35% hospitals/labs, 7% annotators),
 - 30% to developers,
 - 28% pooled for platform sustainability.
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4. Medical Impact

4.1 Prostate Cancer Diagnostics

- **Error Reduction:** Synthetic-trained AI reduced diagnostic errors by **32%** in Nigeria (1,500 patients) vs. traditional workflows **[10]**.
- **Privacy Assurance:** No raw patient data left hospital servers; annotations used synthetic NFTs only.

4.2 Endometrial Cancer Workflow

- **Turnaround Time:** Synthetic-augmented AI cut report delays from 21 days to **48 hours** in India **[11]**.
 - **Bias Mitigation:** Synthetic datasets balanced for ethnic diversity (e.g., 40% Asian, 30% African cases) reduced subtype misclassification by 25% **[12]**.
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5. Roadmap

Year	Milestone	Key Metric
2025	Launch prostate/endometrial AI pilots	95% sensitivity on synthetic-to-real validation.
2026	Scale synthetic pipelines to 20+ hospitals/labs	50% cost reduction in synthetic data generation.
2027	Expand to lung, cervical cancers	10,000+ synthetic NFT datasets.
2030	Global scale: 1,000+ hospitals/labs	10+ cancer diagnostic AIs

6. Ethics & Compliance

- **Privacy Guarantee:** Raw patient data never leaves hospitals; synthetic NFTs are non-reversible to original data [13].
- **Bias Audits:** Synthetic datasets engineered to match WHO demographic guidelines (age, sex, ethnicity) [14].
- **Regulatory Alignment:** Complies with GDPR, HIPAA, and LMIC data sovereignty laws via synthetic-first design [15].

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