

## Title

AI-ENHANCED SPECTROSCOPIC ANALYSIS OF DRIED BLOOD SPOTS FOR MULTI-ANALYTE DETECTION AND QUALITY-CONTROL REDUNDANCY

## Description

### Technical Field

The disclosure pertains to point-of-care and laboratory diagnostics, specifically to **data-driven interpretation of mid-infrared (mid-IR) spectra** acquired from **dried blood spot (DBS)** specimens to generate **binary disease status outputs** for an extensible library of analytes including infectious, genetic, metabolic, toxicological, oncologic, autoimmune, hormonal, nutritional, inflammatory and forensic markers.

### Background

Conventional DBS assays typically rely on tandem MS or immunoassays that are analyte-specific, costly, and cold-chain-dependent. Mid-IR spectroscopy captures a holistic biomolecular fingerprint ( $\approx 451$  latent variables in our calibration transfer matrix). Previous work (ibid) demonstrated automated PID detection from DBS. There is an unmet need to **extend the same single-scan workflow** to a comprehensive panel spanning  $>100$  clinically relevant conditions, while also providing **built-in redundancy for parallel rapid-test workflows (e.g., HIV serology QC)**.

### Summary of the Disclosure

1. **Single-Scan, Multi-Analyte Pipeline** — A handheld ATR-FTIR reader collects a 30 s spectrum ( $4000-650\text{ cm}^{-1}$ ). A convolutional-autoencoder compresses the spectrum into a 451-element latent vector (L1). A modular ensemble of binary classifiers  $\{C_1, \dots, C_n\}$  maps L1 to  $n$  yes/no outputs where  $n \geq 6$  by default and is extensible.
2. **Analyte Library (v5.0)** — Table 1 lists the default 35 condition clusters grouped into six umbrella categories (Infectious, Hemoglobinopathies, Inborn Metabolic Errors, Chronic Metabolic Risk, Oncologic/Auto-Immune, Toxicological & Exposure). Each classifier can be swapped or re-weighted without re-acquiring spectra.
3. **Quality-Control Redundancy** — A rules engine flags discordance between the spectrum-based HIV classifier and any external serology recorded in HL7 field OBX-17, enabling automated recall for NAAT confirmatory testing.
4. **Cloud/LIMS Integration** — All raw spectra, latent vectors, and binary outputs are persisted in SuperSoft™ LIMS; audit hashes are written to Hyperledger Fabric.

5. **Portable Cartridge & Card Design** — A disposable ATR card embeds a QR code linking the blood spot to patient metadata; hydrophobic stencil confines 3  $\mu\text{L}$ . No cold chain required  $<40^\circ\text{C}$  for 21 days.
6. **Sensitivity/Specificity** — Blind hold-out validation across 6,400 clinical DBS shows 88–94 % AUROC per analyte (Table 2); continuous retraining is supported via federated learning.

## Brief Description of the Drawings

- **Fig. 1** Work-flow swim-lane from finger-stick to binary report.
- **Fig. 2** Neural architecture: 1024-x1 input, encoder bottleneck (451 dims), analytic plug-in classifiers.
- **Fig. 3** ATR card exploded view with QR/UID looping to LIMS.
- **Fig. 4** QC discordance dashboard highlighting HIV rapid-test false-negative catch.
- **Fig. 5** Extensibility diagram: adding Dengue classifier via transfer learning.

## Detailed Description

### 1. Specimen Collection & Preparation

A  $3.0 \pm 0.5 \mu\text{L}$  capillary drop from a standard finger-stick is deposited onto an ATR-card capture zone ( $d=6 \text{ mm}$ ). The spot is air-dried  $\leq 10 \text{ min}$  and placed face-down on a diamond ATR crystal integrated into a handheld FTIR engine (Michelson interferometer with DLATGS detector,  $4 \text{ cm}^{-1}$  resolution, 8 scans averaged).

### 2. Spectral Acquisition & Pre-Processing

Raw interferograms are Fourier-transformed; water-vapour compensation and rubber-band baseline correction are applied. VSN-norm and second-derivative Savitzky-Golay filtering (15-pt, 2<sup>nd</sup> order) yield a feature-stable spectrum. The dimensionality-reduction autoencoder is pre-trained on  $>50 \text{ k}$  reference spectra spanning age, sex and ethnic variability.

### 3. Modular Classifier Library

Each classifier  $C_i$  implements either (a) logistic regression on selected latent variables, (b) gradient-boosted decision trees, or (c) a 3-layer perceptron; the interface exposes `predict()`  $\rightarrow \{0, 1, \text{prob}\}$ .

#### Default Condition Clusters (Table 1)

ID	Umbrella Category	Representative Conditions / Analytes	Binary Output Field	Hold-out AUROC
INF-HIV	Infectious – Viral	HIV-1 & HIV-2 acute/chronic infection	HIV_POSSIBLE	0.93

INF-MAL	Infectious – Parasitic	<i>Plasmodium falciparum</i> , <i>P. vivax</i> malaria	MAL_POSSIBLE	0.92
INF-HBV	Infectious – Viral	Hepatitis B virus (HBsAg-positive)	HBV_POSSIBLE	0.91
INF-HCV	Infectious – Viral	Hepatitis C virus	HCV_POSSIBLE	0.90
INF-DEN	Infectious – Viral	Dengue virus	DEN_POSSIBLE	0.90
INF-TB	Infectious – Bacterial	Active pulmonary tuberculosis	TB_POSSIBLE	0.89
HBPD-SC	Hemoglobinopathies	Sickle-cell diseases	SCD_FLAG	0.94
HBPD-βT	Hemoglobinopathies	β-Thalassemia major/minor	BTHAL_FLAG	0.93
IME-PKU	Inborn Errors	Phenylketonuria	PKU_FLAG	0.94
IME-MSUD	Inborn Errors	Maple-syrup-urine disease	MSUD_FLAG	0.90
IME-MCAD	Inborn Errors	Medium-chain acyl-CoA dehydrogenase deficiency	MCAD_FLAG	0.88
CMR-T2D	Chronic Metabolic	Type-2 diabetes risk	T2D_RISK	0.90
CMR-DYS	Chronic Metabolic	Dyslipidaemia (high LDL/low HDL)	DYS_LIPID	0.91
ONC-GI	Oncologic	Early GI cancers	ONC_GI_FLAG	0.91
ONC-BR	Oncologic	Early-stage breast cancer	ONC_BR_FLAG	0.90
ONC-BRAIN	Oncologic	Primary brain tumors	ONC_BRAIN_FLAG	0.93
AUTO-RA	Auto-Immune	Rheumatoid arthritis	AUTO_RA_FLAG	0.88
AUTO-AS	Auto-Immune	Ankylosing spondylitis	AUTO_AS_FLAG	0.88

TOX-METH	Toxicology	Methamphetamine exposure	METH_FLAG	0.87*
TOX-OPI	Toxicology	Chronic opioid exposure	OPIOID_FLAG	0.86*
TOX-PB	Environmental Exposure	Lead (Pb) toxicity signature	LEAD_FLAG	0.85*
FOR-AGE	Forensic	Bloodstain age ≤48 h vs >48 h	AGE_FLAG	0.92
FOR-SPEC	Forensic	Human vs. non-human blood	SPECIES_FLAG	0.98

*Prototype models trained on spiked or simulated datasets; AUROC expected to improve with pilot data.*

(Table 1)

Cluster ID	Representative Conditions	Output Field	AUROC (hold-out)
INF-HIV	HIV-1/2 infection	HIV\_POSSIBLE	0.93
INF-MAL	Malaria (P. falciparum/vivax)	MAL\_POSSIBLE	0.92
INF-HBV	Hepatitis B	HBV\_POSSIBLE	0.91
HBPD-SC	Sickle-cell disease/trait	SCD\_FLAG	0.94
HBPD-βT	β-Thalassemia major/minor	BTHAL\_FLAG	0.93
IME-PKU	Phenylketonuria	PKU\_FLAG	0.94
IME-MSUD	Maple-syrup-urine disease	MSUD\_FLAG	0.90
CMR-T2D	Type-2 diabetes / dysglycaemia	T2D\_RISK	0.90
ONC-GI	Early GI cancers (liver/gastric/CRC)	ONC\_GI\_FLAG	0.91
ONC-AUTO	Ankylosing spondylitis / RA / OA	AUTO\_FLAG	0.88
TOX-METH	Methamphetamine exposure	METH\_FLAG	0.87<sup>b</sup>
...	...	...	...

<sup>b</sup> prototype model trained on spiked DBS.

#### 4. QC Discordance Engine

If HIV\_RAPID\_RESULT==NEG AND HIV\_POSSIBLE==1 then system generates QC\_ALERT with reason code DISCORDANT\_HIV, prompting reflex NAAT or repeat testing.

#### 5. Extensibility & Model Update

New analytes are added by:

1. Uploading ≥120 labeled spectra to the secure model-training bucket.

2. Running the AutoML pipeline; candidate models with cross-val AUROC>0.85 are provisionally accepted.
3. Model artefacts (.onnx) are versioned and deployed via Canary release.

## 6. Example Use Cases

- **Pediatric HIV Program** — AHF clinics embed the card after standard rapid test; QC alerts led to 2 additional HIV cases per 1,000 screened children in a Tanzanian pilot.
- **Neonatal DBS Screening** — State-lab applies assay to routine heel-prick cards; PKU, MSUD and  $\beta$ -thalassemia detected in a single scan.
- **Forensic Field Kit** — Crime-scene investigators identify species origin and estimate age of bloodstain; IR-derived PMI  $\pm 3$  h up to 48 h.

## 7. Hardware Embodiments

- **Portable Handheld** — 750 g device with Li-ion battery (6 h runtime). Wi-Fi/LoRa telemetry; optional solar charger.
- **Benchtop 96-Well ATR Array** — High-throughput head processes Guthrie cards without manual punching; robotic XY stage aligns spots.

## 8. Software & Security

All patient-linked data are AES-256 encrypted in transit and at rest; audit logs immutably persisted via Hyperledger Fabric channel `ftirdbs-audit`. FHIR-compatible API exposes results to EMR.