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## Introduction

Exhaled Breath (EB) contains numerous bioproducts that act as biomarkers and reflect well-being. Biomarkers represent the physiological and enzyme reactions occurring in the body, and their quantification in EB can help detect disorders.<sup>1</sup> Compared to usual clinical tests, breath monitoring is a non-invasive method; hence, the chances of infection are less and suitable for long-term clinical monitoring.<sup>2,3</sup>

Linus Pauling first proposed breath detection in 1970 and successfully detected around 250 bioproducts in the EB by gas chromatography. Breath analysis is done by either monitoring EB in the gas phase or tracking it in the aqueous phase.<sup>4,5</sup> Hydrogen peroxide (H<sub>2</sub>O<sub>2</sub>) is an excellent example and one of the crucial markers in treating numerous diseases.<sup>6</sup> The EB of people with asthma, systemic inflammation, chronic obstructive pulmonary disease (COPD), acute respiratory distress syndrome (ARDS), bronchiectasis, systemic sclerosis, cystic View Journal | View Issue

# A smartphone-interfaced, low-cost colorimetry biosensor for selective detection of bronchiectasis *via* an artificial neural network

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Exhaled breath (EB) contains several macromolecules that can be exploited as biomarkers to provide clinical information about various diseases. Hydrogen peroxide ( $H_2O_2$ ) is a biomarker because it indicates bronchiectasis in humans. This paper presents a non-invasive, low-cost, and portable quantitative analysis for monitoring and quantifying  $H_2O_2$  in EB. The sensing unit works on colorimetry by the synergetic effect of eosin blue, potassium permanganate, and starch-iodine (EPS) systems. Various sampling conditions like pH, response time, concentration, temperature and selectivity were examined. The UV-vis absorption study of the assay showed that the dye system could detect as low as ~0.011 ppm levels of  $H_2O_2$ . A smart device-assisted detection unit that rapidly detects red, green and blue (RGB) values has been interfaced for practical and real-time application. The RGB value-based quantification of the  $H_2O_2$  level was calibrated against NMR spectroscopy and exhibited a close correlation. Further, we adopted a machine learning approach to predict  $H_2O_2$  concentration. For the evaluation, an artificial neural network (ANN) regression model returned 0.941  $R^2$  suggesting its great prospect for discrete level quantification of  $H_2O_2$ . The outcomes exemplified that the sensor could be used to detect bronchiectasis from exhaled breath.

fibrosis, systemic inflammation, uremia, and pneumonia have elevated levels of  $H_2O_2$  (see Table 1). The  $H_2O_2$  levels for these diseases vary between 10 nmol  $L^{-1}$  to 10 µmol  $L^{-1}$ .<sup>7</sup>

Bronchiectasis is a chronic inflammatory lung illness marked by permanent bronchial dilation. Airway secretions include high quantities of proinflammatory cytokines, and neutrophils are the most common cells in the airway lumen. Bronchial damage occurs in people with bronchiectasis due to neutrophil inflammatory agents generated in response to bacterial infection.<sup>8</sup>

Inflammatory cells such as neutrophils, eosinophils, and activated macrophages produce a lot of superoxide anion  $(O_2^{-})$ , which is disputed from hydrogen peroxide either spontaneously or with the help of enzymes.  $H_2O_2$  appears to be a significant reactive oxygen species that cause cellular harm and produces other reactive oxygen species such as hydroxyl radicals and lipid peroxidation products.<sup>9</sup>

Furthermore,  $H_2O_2$  is a key compound in the food industry, chemical industry, laboratory, environmental, and pharmaceutical analysis.<sup>10</sup> Conventional techniques to quantify  $H_2O_2$ include spectrophotometer, chemiluminescence, electrochemistry, spectro fluorometry, and surface plasmon resonance. These techniques are of high cost and require heavy-duty instruments. Even though enzyme-based electrochemistry has high selectivity and sensitivity, the difficulty in immobilizing and stabilizing enzymes limits their use.<sup>11-13</sup>

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Table 1 Biomarkers detected in EB correlate with various pulmonary diseases.<sup>20</sup>

| Diseases                                      | Biomarkers   |  |
|---|--|--|
| Smoking                                       | 8-Isoprostane, $H_2O_2$  |  |
| Chronic obstructive pulmonary disease         | Serotonin, $H_2O_2$ , 8-isoprostane, cytokines   |  |
| Asthma  | Leukotrienes, H <sub>2</sub> O <sub>2</sub> , 8-isoprostane, nitrotyrosine thiobarbituric acid-reactive products |  |
| Bronchiectasis                                | $H_2O_2$   |  |
| Chronic bronchitis                            | Leukotrienes   |  |
| Cystic fibrosis/idiopathic pulmonary fibrosis | Nitrite, IL-8, H <sub>2</sub> O <sub>2</sub> , 8-isoprostane   |  |
| Acute respiratory distress syndrome           | $H_2O_2$ , 8-isoprostane, PGE <sub>2</sub>   |  |

Jeong-Hyeop *et al.* suggested a biosensor that works by the electrocatalytic activity of horseradish peroxidase (HRP)encapsulated protein nanoparticles (HEPNP) for the detection of  $H_2O_2$ .<sup>14</sup> Ali-saad Elewi *et al.* also developed a biosensor based on hemoglobin immobilized on a screen-printed carbon electrode (SPCE).<sup>15</sup> Despite these advancements, there is still a pressing need to create simple, cost-effective, non-invasive, and quick  $H_2O_2$  quantification methods.<sup>16</sup>

Among all the analytical procedures, the colorimetric assay has shown tremendous attention due to its simplicity, speed of execution, high sensitivity and selectivity. Additionally, the colorimetric imaging device overcomes these problems of uneconomically tedious and complex procedures. Further, a smoothly miniaturized alternative was developed where light is separated into a set of color spaces, primarily red, green, and blue (RGB).<sup>17</sup> However, color space-assisted colorimetric analysis significantly impacts ambient light conditions and camera optics. This can be addressed by controlling and confining the light conditions while characterizing the sample.

Further, advanced algorithms such as machine learning can be adopted for statistical analysis. With its powerful utilities like self-learning from the data and automated decisionmaking, machine learning is a dominant method for quantitative evaluation. Combining machine learning with colorimetry will enhance its adaptability and flexibility to emerging frameworks such as Internet of Things (IoT) systems.<sup>18,19</sup>

This study has adopted a three-dye system for more accurate and precise colorimetry sensing. An IoT-enabled detection unit with controlled light conditions can detect RGB color space values within one minute and impart supplementary  $H_2O_2$  data. The detection results are displayed on mobile with designated RGB parameters correspond to the specific concentration of the  $\rm H_2O_2$ . Further, we have explored the machine learning algorithm to estimate  $\rm H_2O_2$  levels based on the RGB value extracted from the sensor prototype. Our research will pave the way for novel, handy, compact, and versatile sensors and, in turn, to identify and analyze  $\rm H_2O_2$  in the exhaled breath, which serves as an essential biomarker for detecting body bronchiectasis.

## Materials and methods

#### Materials and instruments

Hydrogen peroxide (30% w/w), acetonitrile, acetone (99.5% w/v), benzene (37% w/v), ammonia (25%), chloroform, deuterium oxide, eosin blue (2% w/v), formaldehyde (99% w/v), methanol, porphyrin solution, potassium iodide, potassium permanganate (99% w/v), silver nanowire, soluble starch powder ( $M_w$ 692.661 g mol<sup>-1</sup>) and toluene (38% w/v) were purchased from Sigma-Aldrich. All the chemicals used in the experiment were of analytical grade.

#### Methods

Starch solution of 0.1 mM was prepared by mixing corn starch powder in double distilled (DI) water. The mixture was heated to dissolve starch completely. Thereafter, a drop of iodine solution was added in 10 mL of 0.1 mM starch solution, that turns the solution in blue color. Further, to enhance the  $H_2O_2$  sensing properties, a drop of silver nanowire was added in the as prepared starch-iodine solution.<sup>21</sup>

The solutions of potassium permanganate (KMnO<sub>4</sub>) and eosin blue were also prepared in DI water at a concentration of 3 mM. The pH effect of the test analyte with eosin blue, KMnO<sub>4</sub> and silver nanowire dispersed starch iodine (SI) dyes was analyzed in acidic (2, 4, 6), neutral, and basic (9, 12) solutions.



Fig. 1 Fabricated smartphone-assisted sensor prototype.

| Dyes              | Dye indication | pH(x), Temperature<br>(y in °C), biomarker (H <sub>2</sub> O <sub>2</sub> (H))-concentration (z in ppm) indication |  |
|-------------------|----------------|--|--|
| KMnO <sub>4</sub> | КМ             | KM(PxTyHz)   |  |
| Eosin blue        | EB             | EB(PxTyHz)   |  |
| Starch-iodine     | SI             | SI(PxTyHz)   |  |

Each dye solution (10 mL) was added with 1 mL of 0.3 ppm  $H_2O_2$  solution in the adjusted pH. Changes in color and corresponding response time were observed. Dyes solutions were then characterised by UV-visible spectroscopy technique. The detection limit of the dyes was monitored in the range of 0.003 to 5 ppm concentrations of  $H_2O_2$  solution and temperature effect was studied. Selectivity analysis was performed using 0.03 ppm of  $H_2O_2$ , formaldehyde, acetone, nitric oxide, ammonia, toluene and benzene at room temperature.

#### Characterization

A Biochrom UV-vis Spectrophotometer (scanning range of 190– 1100 nm) was used for characterization. Here a scanning range of 300–750 nm with a medium scan speed was adapted. Bruker Advance III 500 MHz spectrometer (Bruker Bio spin, Rheinstetten, Germany) with a BBFOPLUS Smart Probe equipped with a Bruker Automatic Sample Changer was used to perform NMR spectroscopy studies (B-ACS 120).

#### IoT-based sensor prototype

For color space-based analysis, a sensor unit was fabricated using the 3D printer (QIDI 3D printer). The sensor prototype offered controlled ambient light conditions and optics during color space-based  $H_2O_2$  quantification. The sensor prototype has three independent regions, *i.e.*, a light source zone, sample zone, and detection zone, as shown in Fig. 1. The light source consists of white light LED. The sample zone with three cuvettes consisting of individual eosin blue, KMnO<sub>4</sub> and SI (EPS) dye solutions are placed sequentially in each section. The light source, sample zone, and detector are aligned horizontally to increase the reliability of the prototype. The light from the LED falls on the cuvettes containing EPS dye solution, and the detector intercepts light transmitted from the sample zone. The detector examines the light that has been transmitted, and light is divided into red, green and blue color spaces. The results are communicated to the screen *via* the Bluetooth technique.

#### Validation of H<sub>2</sub>O<sub>2</sub> measurements

To evaluate the applicability of the proposed sensor prototype, comparisons were made concerning heavy-duty instruments like the NMR spectrometer.4 The solution employed in the NMR experiments of  $H_2O_2$  breakdown contains 75% (v/v) acetonitrile, and the rest is methanol. 3.1 mL of solvent, 5.0 mL of H<sub>2</sub>O<sub>2</sub>, and 10 mL of chloroform were combined in a 5 mL reaction vial, and 463 mL of this mixture was placed in a precision NMR tube. Before starting an experiment, an NMR tube without the catalyst was placed in the NMR spectrometer to lock the field and shim the magnet. After the NMR spectrometer was set up, 37 L of a 1.0 mM porphyrin solution was injected into the reaction mixture's NMR tube. An NMR tube containing 50 percent deuterium oxide and 50 percent water was shaken to mix the sample. The spectrometer was then allowed to lock onto the sample after the NMR tube containing the reacting sample was inserted.

## **Results and discussion**

To fabricate the extremely sensitive colorimetric VOC sensor,  $H_2O_2$  solution (concentration ranging from 0.003–5 ppm) was added to the dye solutions in an acidic, basic, and neutral



Fig. 2 SI dye response toward  $H_2O_2$ . (a) pH adjusted SI dye solution before and after adding 0.3 ppm  $H_2O_2$  solution at room temperature. (b) (b) Absorbance spectra of neutral SI dye solution before and after adding 0.3 ppm  $H_2O_2$ .





media. Any obvious color changes were monitored. Response time, pH impact, temperature effect, concentration effect, and the dyes' selective nature were also investigated and assessed.

Table 2 represents the list of notations for the study. Here x stands for specific pH, y stands for a specific temperature, and z stands for biomarker concentration.

#### Response time and pH effect

 $H_2O_2$  was detected using SI, eosin blue, and KMnO<sub>4</sub> dye solutions with pH 2, 4, 6, 7, 9, and 12. At room temperature, 0.3 ppm of  $H_2O_2$  in 1 mL DI water was added to the dye's solution. The dyes' response time was calculated by noting the time between the addition of  $H_2O_2$  and the visible color change noticed in the dye's solution. In  $H_2O_2$  detection analyses using the SI dye, the color shift was visible in neutral and basic solutions but not in the acidic media. The color shift in the SI dye solution after adding 0.3 ppm  $H_2O_2$  is shown in Fig. 2a. From Fig. 2a, it can be inferred that the neutral and basic starch dye solution color tends to diminish and turns slight blue with  $H_2O_2$ . Herein, irrespective of pH value, starch dye response time decreases with increased  $H_2O_2$  concentration. In all the pH-adjusted

solutions, the estimated response time for 5, 3, 0.3, and 0.03 ppm H<sub>2</sub>O<sub>2</sub> was nearly 6 min, 8 min 22 s, 12 min 16 s, and 20 min, respectively. It can be inferred that the redox reaction between potassium iodide and H2O2 followed by the color reaction of SI could be finished within 20 min for H<sub>2</sub>O<sub>2</sub> concentration as low as 0.03 ppm. In the acidic medium more number of H<sup>+</sup> ions are available, as a result most of the iodine ions are converted to the iodine followed by the formation of the complex compound that gives the blue color. Thus the addition of H<sub>2</sub>O<sub>2</sub> in the acidic dye medium only intensify the blue color and not much change visible color change is observed. However, with increase in the pH from neutral to basic the H<sup>+</sup> ion concentration decreases which decreases the formation of the iodine ions and formation of complex compound respectively. As a result, faint pink color of solution is observed. Thus, addition of H<sub>2</sub>O<sub>2</sub> catalysis the formation of the iodine in the neutral and basic medium that results in the appearance of slight blue color. Eqn (1) and (2) gives the kinetics of the response of the SI dye solution system and H2O2.

$$H_2O_2 + 2I^- + 2H^+ \rightarrow I_2 + 2H_2$$
 (1)



Fig. 4 KMnO<sub>4</sub> dye response toward  $H_2O_2$ . (a) pH adjusted KMnO<sub>4</sub> dye solution before and after adding 0.3 ppm  $H_2O_2$  solution at room temperature. (b) Absorbance spectra of neutral KMnO<sub>4</sub> dye solution before and after adding 0.3 ppm  $H_2O_2$  solution.





Fig. 5 Dyes sensitivity towards  $H_2O_2$  detection. (a and b) UV-vis absorption curve with change in  $H_2O_2$  concentration and corresponding calibration plot of eosin blue dye, respectively. (c and d) UV-vis absorption curve with  $H_2O_2$  concentration shift and corresponding calibration plot of KMnO<sub>4</sub> dye, respectively. (e and f) UV-vis absorption curve with varying  $H_2O_2$  concentration and corresponding calibration plot of starch dye, respectively.

$$nI_2 + 6n(C_6H_{10}O_5) \rightarrow 2n(C_{18}H_{30}O_{15}I)$$
 (2)

Here, H  $_2O_2$  liberated iodine from potassium iodide (eqn (1)), followed by the reaction of starch and iodine, resulting in a complex classic response (eqn (2)).<sup>22</sup>

For the  $H_2O_2$  assay in eosin blue solution, the visible color change was noticed in pH 12 solution for all  $H_2O_2$  concentrations, *i.e.*, 0.03–5 ppm. Fig. 3a shows the color change in the EB solution after adding 0.3 ppm  $H_2O_2$ . An apparent color change from red to orange is observed in the dye solution of pH 12. The corresponding absorbance curve of pH solution before and after adding  $H_2O_2$  is shown in Fig. 3b. The response time of pH 12 solution increased with a decrease in the concentration of  $H_2O_2$  and was estimated to be 6 min, 8 min, 15 min, and 25 min for 5, 3, 0.3, 0.03 ppm test solution, respectively. In eosin blue dye, the presence of polycyclic aromatic ring produces the most intense peak at ~518 nm. Therefore, change of color indicates the degradation of the polycyclic aromatic rings. In neutral and basic aqueous solution, eosin blue (EB) exist as  $EB^{2-}(aq)$ , while in week and strong acidic medium eosin blue forms  $HEB^{-}(aq)$  and  $H_2EB$  (aq), respectively. Thus, with the formation of ionic compound, there are more chances of eosin blue degradation in

Paper



Fig. 6 Temperature effect on  $H_2O_2$  detection in EPS dye solutions system.

week acidic, neutral and basic medium. Moreover,  $H_2O_2$  is more stable in the acidic medium compared to the basic medium. Thus, in acid medium  $H_2O_2$  doesn't undergo auto degradation to produce free radical that can react with EB and perform dye degradation. Whereas, in the pH 12 solution,  $H_2O_2$  undergoes auto degradation to form HO' radical, as given by eqn (3) and eqn (4).

$$H_2O_2 \rightarrow 2H^+ + HO_2^- \tag{3}$$

$$H_2O_2 + HO_2^- \rightarrow HO_2^{\cdot} + HO^{\cdot} + OH^-$$
(4)

The generated HO' radicals react with eosin blue and results in the degradation of the dye that is represented by the degradation of the dye color in pH 12 dye solution, as shown in Fig. 3a.

Interestingly, the  $H_2O_2$  assay in KMnO<sub>4</sub> dye solution exhibited a prominently visible color change in all pH mediums compared to SI and EB dye solutions. In addition, the color variation of the solution from violet to light orange and violet to deep orange after adding  $H_2O_2$  is visually observed irrespective of the concentration of the test solution. The color change in the KMnO<sub>4</sub> solution after adding 0.3 ppm  $H_2O_2$  is shown in Fig. 4a. At concentrations of  $H_2O_2$  as low as 0.3 ppm, the apparent color change may be identified, providing a simple method for detecting  $H_2O_2$  with the naked eye.

Like other dyes, the response time of  $\text{KMnO}_4$  increases with a decrease in the test solution concentration. Compared to the acidic and basic medium, the neutral  $\text{KMnO}_4$  solution exhibits high assay performance with a 1, 7, 13, and 27 s response time for 5, 3, 0.3, and 0.03 ppm, respectively. Redox reaction of  $\text{KMnO}_4$  is given by eqn (5).

$$5H_2O_2 + 6H^+ + 2KMnO_4 \rightarrow 5O_2 + 2Mn^{2+} + 8H_2O + 2K^+$$
 (5)

Further in the study, for SI and KMnO<sub>4</sub> dye solution we have opted for the neutral dye medium, since it removes the pH adjustment step and simplifies application of the dye in the fabricated sensing prototype system.

#### Concentration effect and limit of detection

The UV-vis study was carried out to determine the sensitivity of the colorimetric EPS system by adjusting the  $H_2O_2$  content in the SI (pH 7), KMnO<sub>4</sub> (pH 7) and EB (pH 12) dye solutions from 0.003 to 5 ppm. Fig. 5 illustrates the UV-vis absorbance plot of the dyes with changes in  $H_2O_2$  from 0.003–5 ppm and the corresponding calibration for determining the dye's LOD towards  $H_2O_2$ . The LOD was estimated using the  $3\sigma/m$  criterion after the calibration curve was plotted using the dye's peak absorbance at a certain wavelength. Where *m* is the calibration plot's slope and  $\sigma$  is the intercept's standard deviation. Fig. 5a and b show the UV-vis plot of the EB dye as a function of  $H_2O_2$  concentration and calibration cure. As the concentration of  $H_2O_2$  rises, the absorbance also rises (Fig. 5a). The calibration curve was plotted using the dye's peak absorbance at 518 nm for various  $H_2O_2$ concentrations. The linear fitting was performed in the range of



Fig. 7 Selectivity analysis of biomarkers in different dye solutions (a) Eosin Blue dye solutions (b) KMnO<sub>4</sub> dye solution (c) SI dye solution.

Table 3 Analysis of H<sub>2</sub>O<sub>2</sub>-EPS mixture using Sensor Prototype

| H <sub>2</sub> O <sub>2</sub> -EPS Mixture | RGB values obtained for different light source |            |             |             |  |
|--|--|------------|-------------|-------------|--|
|  | Red light                                      | Blue light | White light | Green light |  |
| EPS  | 255,0,0  | 103,0,255  | 255,0,226   | 19,88,50    |  |
| $E(H_2O_2) PS$                             | 116,0,8  | 33,0,109   | 124,0,109   | 22,44,7     |  |
| $EP(H_2O_2)$ S                             | 103,0,5  | 0,38,90    | 82,94,101   | 0,94,4      |  |
| $EPS(H_2O_2)$                              | 179,0,9  | 36,0,255   | 175,47,239  | 0,87,39     |  |
| $E(H_2O_2) P(H_2O_2) S$                    | 175,0,3  | 32,50,109  | 165,118,98  | 0,130,0     |  |
| $E(H_2O_2) PS(H_2O_2)$                     | 181,07   | 42,6,197   | 181,56,184  | 0,89,30     |  |
| $EP(H_2O_2) S(H_2O_2)$                     | 123,0,5  | 20,34,84   | 124,92,76   | 0,104,0     |  |
| $E(H_2O_2) P(H_2O_2) S(H_2O_2)$            | 121,0,3  | 27,34,82   | 116,90,73   | 0,97,0      |  |



Fig. 8 3D plot, representing RGB value corresponding to the  $\rm H_2O_2$  level.



Fig. 9 Comparison of NMR data and colorimetric data for quantification of  ${\rm H}_2{\rm O}_2$ 

0.003–5 ppm H<sub>2</sub>O<sub>2</sub> and estimated LOD of EB was ~0.011 ppm (Fig. 5b),  $y = (2.3984) x + (1.57804 \pm 0.00906)$ ;  $R^2 = 0.94159$ . Like EB, an increase in absorbance with an increase in H<sub>2</sub>O<sub>2</sub> ppm

level was noticed for KMnO<sub>4</sub> dye (see Fig. 5c). The calibration curve was plotted from 0.003-0.3 ppm concentration by considering the absorbance at 525 nm. Fig. 5d shows the calibration curve of KMnO<sub>4</sub> dye in the linear range of 0.003-5 ppm. The linear fit to the data revealed an LOD of  $\sim 0.025$  ppm towards  $H_2O_2$  sensing by KMnO<sub>4</sub> dye,  $y = (0.12484) x + (0.6571 \pm$ 0.001044);  $R^2 = 0.98944$ ). SI dye also exhibited a linear relationship with the concentration, as shown in Fig. 5e. With an increase in H<sub>2</sub>O<sub>2</sub> concentration, the absorbance also increased. The calibration curve was plotted using the peak dye absorbance at 354 nm for various H<sub>2</sub>O<sub>2</sub>. The linear fitting was done in the range of 0.003-5 ppm H<sub>2</sub>O<sub>2</sub> and the predicted LOD of the starch dye was 0.044 ppm,  $y = (0.0416) x + (0.03035 6.109 \times$  $10^{-4}$ );  $R^2 = 0.9906$ ). According to the sensitivity study, the EPS dye system has a high sensitivity to H<sub>2</sub>O<sub>2</sub>, with a detection limit of 0.011 ppm.

#### **Temperature effect**

Even though they are critical, shelf life and stability qualities are frequently understudied or ignored in the literature. At an elevated temperature, instability or aging can be hastened. To investigate this parameter, dye solutions were heated at different temperatures, *i.e.*, 25 °C, 50 °C, 75 °C and 100 °C. The dye solutions were given 1 mL of the test solution, with a concentration of 0.03 ppm. As demonstrated in Fig. 6, the absorption intensity of the EPS dye solution remained nearly constant regardless of temperature change. This showed that the EPS dye system is temperature stable, which is important for the colorimetric sensor.

#### Selectivity analysis

Control studies with potential interfering biomarkers in breath such as acetone, benzene, ammonia, formaldehyde, toluene, and nitric oxide were done in EPS dye solution to evaluate the selectivity of  $H_2O_2$  detection. The  $H_2O_2$  and possible interfering analytes concentrations were both 0.03 ppm. The dye's selectivity for  $H_2O_2$  was determined using UV-vis analysis and eqn (6).

$$\Delta \lambda = \frac{\lambda_{\rm X} - \lambda_0}{\lambda_0} \times 100 \tag{6}$$

#### Paper



Fig. 10 (a) Flow diagram of regression analysis using artificial neural network (ANN) regression algorithm. (b) The ANN machine learning algorithm plot shows the linear fit between the target and output value (left) and the change in color of the colorimetric dye solution after detecting  $H_2O_2$  in the range of 0.1 to 200 ppm level (right).

where  $\lambda_x$  is the analyte's unique peak absorbance wavelength, and  $\lambda_0$  is the wavelength of the blank solution with high absorbance. The  $\lambda_x$  value is measured at pH 12 for EB and a neutral solution for KMnO<sub>4</sub> and SI dye. In the case of all three dyes, it was observed that only H<sub>2</sub>O<sub>2</sub> could induce a shift in UVvis peak absorbance, as shown in Fig. 7. According to these findings, other interfering compounds did not compete with the H<sub>2</sub>O<sub>2</sub> chemo-indicator for colorimetric detection. As a result, the EPS dye system can be used to create a highly specific colorimetric sensor for H<sub>2</sub>O<sub>2</sub> molecules.

#### Real-time application of the proposed sensor and evaluation

A portable prototype device with full functions for detecting hydrogen peroxide was constructed. Employing the three dyes as sensing elements, the sensor prototype showed a unique set of RGB values upon exposure to 0.03 ppm hydrogen peroxide (Table 3). Employing the EPS dye system, the sensor prototype displayed a unique set of RGB values when exposed to diverse  $H_2O_2$ concentrations ranging from 0.001 to 200 ppm. The sensor prototype displayed a unique set of RGB values. Fig. 8 shows the 3D plot, representing the RGB value corresponding to the  $H_2O_2$ level. A decrease in the intensity of the 'B' color space value is observed with an increase in the  $H_2O_2$  concentration.

Each RGB value corresponds to a unique concentration of  $H_2O_2$  that can be correlated to clinical information for quantifying bronchiectasis in the human body.

To investigate the accuracy of the proposed colorimetric method, the sensor device was calibrated against the NMR spectroscopy technique. Fig. 9 compares the efficiency of NMR and colorimetric data for  $H_2O_2$  detection. The quantification data at the potential range of 0.5–0.8 V shows the concentration in the range of 0.5–7.2 ppm in NMR and 0.3–7.15 ppm in the sensor prototype. The colorimetry and NMR integration results were mostly in accord, with some overestimated related

compounds due to peak overlap, integral errors, and a lower estimate for the target chemical. The study shows that this innovative method for colorimetric detection and quantification of  $H_2O_2$  is accurate for chemical analysis. Moreover, according to the cost comparison between NMR and colorimetry, it was shown that this IoT coupled sensing prototype is cost-effective as opposed to the high-cost NMR analysis. Furthermore, unlike NMR, this colorimetric technique would yield results in seconds rather than hours.

#### Artificial neural network (ANN) regression

Further, we have also analyzed the RGB color space value extracted from the sample under white light illumination and employed artificial neural network (ANN) regression machine learning algorithms to identify the corresponding H<sub>2</sub>O<sub>2</sub> level. We used 15 different concentrations of H<sub>2</sub>O<sub>2</sub> ranging from 0.001 ppm to 200 ppm at different temperatures of 25  $^\circ C$  and 50 °C. A total of thirty sets of readings were obtained, out of which twenty-five were used for training the model, and five were used for testing. Fig. 10 shows the ANN machine learning algorithm's output results and the color change of colorimetric dye solution detecting H<sub>2</sub>O<sub>2</sub> at 0.1 to 200 ppm-level. Fig. 10a represents the flow diagram of the ANN machine learning algorithm. The algorithm is divided into sub-module layers of the neurons where the neuron gets the H<sub>2</sub>O<sub>2</sub> concentration and RBG value as three input training parameters. The output of the three neurons in the sub-modules is given as input to the neuron in another submodule layer. The output from the neuron is then used for the concentration mapping, and the final output gives the concentration of the H<sub>2</sub>O<sub>2</sub> at the ppmlevel. Fig. 10b shows the ANN machine learning algorithm plot with a linear fit between the target and the output value and a change in color of colorimetric dye solution after detecting H<sub>2</sub>O<sub>2</sub> in the range of 0.1 to 200 ppm-level. The ANN model exhibited ~94% accuracy, indicating that the proposed method fused with machine learning is a great prospect for discrete level quantification of H2O2. The results demonstrate the colorimetry's potential for detecting Volatile Organic Compounds (VOCs) in our exhaled breath. It can also be used in the biomedical, nuclear, food, etc.

## Conclusion

This study examined a new implementation of a colorimetrybased non-invasive, low-cost detection method to classify hydrogen peroxide  $(H_2O_2)$  levels. A portable 3D printed sensing device offering controlled light and optic conditions were designed for detection. Three dyes eosin blue, potassium permanganate (KMnO<sub>4</sub>) and starch-iodine having Silver Nanowire (SI), were used as sensing elements. All the dyes exhibited promising results in detecting  $H_2O_2$ , and the dye samples were characterized by UV-vis spectroscopy. The estimated detection limit for eosin blue, KMnO<sub>4</sub> and SI dye solution was 0.011 ppm, 0.025 ppm and 0.044 ppm, respectively. The dyes apprehended high selectivity for determining  $H_2O_2$  in the presence of other interfering biomarkers. Using the EPS dye system, the Further, a machine learning classifier was trained using the RGB value obtained from the sensor prototype under white light illumination. This improved the robustness of the detection platform, and the adopted algorithm displayed ~94% accuracy, indicating that the proposed colorimetric method fused with machine learning is a great prospect for discrete level quantification. Moreover, the reported sensor device can be versatilely extended to quantify various targets as long as suitable colorimetric agents are available. The successful applications of this novel approach have great potential as a visual sensor platform in the biomedical field, food industry, chemical industry, and environmental and pharmaceutical analysis.

## **Author Contributions**

Mizaj Shabil Sha: synthesis, experimentations, real sample analysis, investigation, data organization and writing – original draft. Muni Raj Maurya: real sample analysis, sensing mechanism interpretation and Writing – review & editing. Muhammad E.H. Chowdhury: writing – Review & Editing. Asan G. A. Muthalif: Software, Validation. Somaya Al-Maadeed: data organization and writing – review & editing. Kishor Kumar Sadasivuni1 (Corresponding author): conceptualization, supervision, writing – review & editing, resources, funding acquisition and project administration.

## Conflicts of interest

The authors declare that they have no conflict of interest.

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## References

- 1 Y. L. Pham and J. Beauchamp, Molecules, 2021, 26, 5514.
- 2 M. R. Maurya, H. Onthath, H. Morsy, N.-U.-S. Riyaz, M. Ibrahim, A. E. Ahmed, R. Abuznad, A. Alruwaili, F. Alsaedi, P. Kasak and K. K. Sadasivuni, *Healthc. Basel Switz.*, 2021, **9**, 1055.
- 3 A. Cervantes Trejo, I. D. Castañeda, A. C. Rodríguez, V. R. Andrade Carmona, M. del P. C. Mercado, L. S. Vale, M. Cruz, S. Barrero Castillero, L. C. Consuelo and M. Di Silvio, *Evid.-Based Complementary Altern. Med.*, 2021, 2021, 5592042.
- 4 M. S. Sha, M. R. Maurya, M. Geetha, B. Kumar, A. M. Abdullah and K. K. Sadasivuni, *Sensors*, 2022, 22, 618.

- 5 M. S. Sha, M. R. Maurya, S. Shafath, J.-J. Cabibihan, A. Al-Ali, R. A. Malik and K. K. Sadasivuni, *ACS Omega*, 2022, 7, 4257– 4266.
- 6 M. Sharifi-Rad, N. V. Anil Kumar, P. Zucca, E. M. Varoni,
  L. Dini, E. Panzarini, J. Rajkovic, P. V. Tsouh Fokou,
  E. Azzini, I. Peluso, A. Prakash Mishra, M. Nigam, Y. El
  Rayess, M. E. Beyrouthy, L. Polito, M. Iriti, N. Martins,
  M. Martorell, A. O. Docea, W. N. Setzer, D. Calina,
  W. C. Cho and J. Sharifi-Rad, *Front. Physiol.*, 2020, 11, 694.
- 7 E. G. Vasiliou, Y. M. Makarovska, I. A. Pneumatikos, N. V. Lolis, E. A. Kalogeratos, E. K. Papadakis and C. A. Georgiou, *J. Braz. Chem. Soc.*, 2007, 18, 1040–1047.
- 8 A. Manfra, J. H. Sharma, J. P. Kilburn, M. Chang and N. Noureddin, *Am. J. Respir. Crit. Care Med.*, 2021, **203**(9).
- 9 M. Jones, D. Neville, J. Gates, T. Jones, C. Fogg, M. Amos-Lyons, M. Chauhan, S. Begum, J. Longstaff, J. Winter, J. Williams, J. Shute, T. Brown, H. Rupani and A. Chauhan, *Eur. Respir. J.*, 2019, DOI: 10.1183/13993003.congress-2019.pa1712.
- 10 R. M. Trujillo, D. E. Barraza, M. L. Zamora, A. Cattani-Scholz and R. E. Madrid, *Sensors*, 2021, 21, 2204.
- 11 X.-S. Chai, Q. Hou, Q. Luo and J. Y. Zhu, *Anal. Chim. Acta*, 2004, **507**, 281–284.

- 12 M.-H. Cheng, A. J. Chicco, D. Ball and T. W. Chen, Sens. Actuators, B, 2022, 360, 131641.
- 13 V. Jain, S. Bhagat and S. Singh, *Sens. Actuators, B*, 2021, 327, 128886.
- 14 J.-H. Shin, M.-J. Lee, J.-H. Choi, J. Song, T.-H. Kim and B.-K. Oh, *Nano Convergence*, 2020, 7, 39.
- 15 A. S. Elewi, S. A. W. Al-Shammaree and A. K. M. A. AL Sammarraie, *Sens. Bio-Sens. Res.*, 2020, **28**, 100340.
- 16 Y. Yu, M. Pan, J. Peng, D. Hu, Y. Hao and Z. Qian, *Chin. Chem. Lett.*, 2022, DOI: 10.1016/j.cclet.2022.02.045.
- 17 L. C. Lopes, A. Santos and P. R. Bueno, *Sens. Actuator A Phys.*, 2022, 4, 100087.
- 18 S. Gruszczyński and W. Gruszczyński, *Geoderma*, 2022, 405, 115451.
- 19 K. Mochida, S. Koda, K. Inoue, T. Hirayama, S. Tanaka, R. Nishii and F. Melgani, *GigaScience*, 2018, **8**, giy153.
- 20 J. Pravda, Mol. Med., 2020, 26, 41.
- 21 S. Koylan, S. Tunca, G. Polat, M. B. Durukan, D. Kim, Y. E. Kalay, S. H. Ko and H. E. Unalan, *Nanoscale*, 2021, 13, 13129–13141.
- 22 Z. Xu, F. Jiang, F. Zhou, Z. Xu, H. Xu and X. Ruan, *Anal. Methods*, 2017, **9**, 3230–3236.