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Ubiquitous Low-Cost Functionalized Multi-Walled Carbon Nanotube Sensors for Distributed Methane Leak Detection

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Abstract—This paper presents a highly sensitive, energy efficient, and low-cost distributed methane (CH$_4$) sensor system (DMSS) for continuous monitoring, detection, and localization of CH$_4$ leaks in natural gas infrastructure, such as transmission and distribution pipelines, wells, and production pads. The CH$_4$ sensing element, a key component of the DMSS, consists of a metal–oxide nanocrystal functionalized multi-walled carbon nanotube mesh, which, in comparison with the existing literature, shows stronger relative resistance change while interacting with lower parts per million concentration of CH$_4$. A Gaussian plume triangulation algorithm has been developed for the DMSS. Given a geometric model of the surrounding environment, the algorithm can precisely detect and localize a CH$_4$ leak as well as estimate its mass emission rate. A UV-based surface recovery technique making the sensor recover ten times faster than the reported ones is presented for the DMSS. A control algorithm based on the UV-accelerated recovery is developed, which facilitates faster leak detection.

Index Terms—Methane, carbon nanotube, chemoresistor.

I. INTRODUCTION

DESPISTE having a shorter lifetime than CO$_2$, methane gas (CH$_4$) is more efficient in trapping radiation [1], [2]. The impact of CH$_4$ on climate change is projected to be 25 times stronger than that of CO$_2$ over a 100-year period [1], [3]. In 2013, 10% of emitted greenhouse gas in the USA was CH$_4$, while 29% of CH$_4$ emission was anthropogenic, i.e., the source is man-made, such as natural gas and petroleum systems [2]. Hence, the need exists for an extensive deployment of low cost, highly sensitive, selective and continuous CH$_4$ monitoring networks throughout the natural gas extraction and distribution infrastructure.

In this paper, we describe a low-cost microfabricated CH$_4$ sensing system and a corresponding detection algorithm for pinpointing leaks in natural gas infrastructures. Our novel Gaussian plume triangulation algorithm can precisely detect and localize a CH$_4$ leak and estimates its mass emission rate (MER), if a geometric model of the surrounding environment is provided. This algorithm complements our novel metal oxide nanocrystals (MONC) functionalized MWCNT mesh based micro-fabricated CH$_4$ sensing element, which is the key constituent for the distributed CH$_4$ sensor system (DMSS) described in this work. A control algorithm for the plume mapping process based on UV-accelerated surface recovery of the functionalized MWCNT CH$_4$ sensor is also described. The reminder of this paper is structured as follows: Section II presents a review on existing methane monitoring technologies; the fabrication process and characterization of the MONC functionalized MWCNT methane sensor are presented in Section III; Section IV explains the triangulation algorithm for remote methane leak detection; a UV-recovery-based plume mapping algorithm is described in Section V; Section VI describes sensor response time and possible interference during outdoor application of the sensor; Section VII summarizes the paper. A preliminary version of this paper was previously presented [4].

II. RELATED WORK

Three types of CH$_4$ sensors have been generally reported in literature: (a) sensors based on filtered infrared (IR) absorption [5], [6], (b) sensors based on Cavity Ring-Down Spectroscopy (CRDS) [3], [7], (c) metal-oxide based chemoresistor/chem-FET [8]–[10].

Filtered IR sensors, which detect CH$_4$ in the ppm range from its absorption of IR light, have been widely used in gas gathering, transmission and distribution pipelines. A number of other hydrocarbon gases also have absorption bands/lines in the same spectral region as CH$_4$, hence, IR detectors are not methane-specific [5]. Also, they are expensive and often need to be cooled to achieve better signal-to-noise ratios, increasing their power consumption [6].

Sensors based on cavity ring-down spectroscopy (CRDS) utilize an optical detection scheme to conduct fast mobile
leak measurements. The isotopic ratios of the stable isotopes of both carbon and hydrogen, along with the atmospheric concentration and distribution of the CH₄, can be used to derive the location and magnitude of CH₄ sources. However, the principal limitation of CRDS is its complex structure and immobility making the detection technology significantly expensive [3], [7].

Metal-oxide chemo-resistor or chem-FET are widely used to detect methane [8]–[10]. Continuous heating is necessary to initiate the surface chemisorption process of oxygen on these metal-oxides. Consequently, the power consumption by these sensors is high (hundreds of mWs and higher) [8]–[10].

CNT based gas sensors have the following properties which are uniquely attributed to their nano-scale dimension: (1) great adsorptive power due to large surface area to volume ratio, (2) stronger modulation of electrical properties (i.e., resistance) upon exposure to target gases, (3) tunable electrical properties correlated with composition and size, and (4) compatibility with low-power micro and nano-electronics [11], [12].

Thus, ppm level of gas detection has been reported using CNT based chemoresistive sensors with a power consumption of only few mWs [13]–[15].

### III. SENSOR FABRICATION AND TESTING

We fabricated MONC functionalized MWCNT CH₄ chemoresistor sensors which are able to detect 10 ppm of CH₄ in dry air at room temperature. A lift-off based photolithography process was implemented to fabricate a set of interdigitated metal (Au/Cr) electrodes on top of SiO₂/Si substrates (Fig. 1a). The gap between the electrodes varied in the range of 5 μm to 10 μm. A solution of MWCNTs/ethanol was drop deposited on the interdigitated electrodes following a baking at 75 °C temperature to remove the solvent. MWCNTs network created a conductive bridge between the two pairs of electrodes (Fig. 1b). The MWCNTs were subsequently surface-activated by O₂ plasma and UV-O₃ for duration of 5 mins to 1 hr. The surface activated MWCNTs network was functionalized with metal oxides such as ZnO [16] or SnO₂ [17] nanocrystals using atomic layer deposition (ALD). Fig. 1c is a transmission electron microscope (TEM) image of a ZnO functionalized MWCNT sample showing uniform distribution of the “spherical” ZnO nanocrystals on the MWCNT surface. Clearly visible lattice fringes of ZnO in the high resolution TEM image suggest high crystal quality of ZnO nanocrystals (Fig. 1d).

The fabricated sensors were tested at room temperature inside an enclosed chamber where 10 ppm CH₄ in dry air mixture was introduced while maintaining a constant (approximately 5 %) relative humidity (RH). The relative resistance change of the sensor \( \Delta R/\Delta R = (R_{CH₄} - R_{air})/R_{air} \) were recorded as a function of time and was found to increase in a monotonic fashion in presence of the CH₄ mixture (Fig. 2).

### IV. TRIANGULATION ALGORITHM FOR REMOTE LEAK DETECTION

In this work, we present a novel Gaussian plume triangulation algorithm which, given a geometric model of the surrounding environment, can precisely detect and localize a CH₄ leak while estimating its mass emission rate. Note that we assumed a faster response sensor comparing to our fabricated MONC-MWCNT sensors. Our current fabrication is focused on reducing the response time of the MONC-MWCNT sensors.

In the proposed approach the MWCNT sensors will be mounted on nodes distributed close to a production pad or pipeline. Fig. 3 illustrates a system where the sensor nodes are distributed on a 10 m X 10 m grid on an area around a natural gas production well head. The approach
relies on varying wind direction to sweep a plume of CH\textsubscript{4} emanating from a nearby leak (e.g. from the well head) across two or more sensors. Once the plume moves across a sensor location due to variations in wind direction, a plume distribution can be reconstructed by using the triangulation detection algorithm which is based on Gaussian plume model.

However, the DMSS can be extended to an arbitrary spatial sensor distribution. The approach relies on varying wind direction to sweep a plume of CH\textsubscript{4} emanating from a nearby leak across two or more sensors. Using the National Wind Technology Center dataset \cite{18} as a baseline (obtained near Boulder, Colorado, US) wind rose plots were analyzed for 4 randomly chosen 18 days period in 2013. The average (prevailing) direction of wind was West or South-West with the wind covering more than 180° in the course of 18 days. This is assumed to be representative wind distribution and is later used by the leak detection algorithm.

\section*{A. Gaussian Plume Model}

A Gaussian plume model \cite{19} is devised for estimating the evolution of a plume emanating from a ground level leak for different wind directions and speeds. For a constant leak rate and a constant wind speed the ground level concentration $\chi (\mu g/m^3)$ at the point $(x,y)$ may be written using the Gaussian plume equation as:

$$\chi(x, y, z_0, H) = \frac{Q \times 10^6}{\pi \sigma_y \sigma_z U} \exp\left[ -\frac{1}{2} \left( \frac{y}{\sigma_y} \right)^2 \right] \times \exp\left[ -\frac{1}{2} \left( \frac{H}{\sigma_z} \right)^2 \right]$$

(1)

where $Q$ (\mu g/s) is the source emission rate, $U$ (m/s) is the wind speed at stack height, $\sigma_y$ and $\sigma_z$ are the standard deviation of concentration distributions in crosswind and vertical directions respectively, $H$ (m) is the effective stack height (the sum of stack height and plume rise), $x$ (m) is the downwind distance from the stack, $y$ (m) is the crosswind distance from the plume centerline, and $z_0$ (m) is the vertical distance from ground level, assumed to be zero.

From the fitting of Pasquill-Gifford-Turner diffusion coefficient curves, $\sigma_y$ and $\sigma_z$ are given as:

$$\sigma_y = cx^d$$

(2)

$$\sigma_z = ax^b$$

(3)

where, constants $a$, $b$, $c$, and $d$ are 0.112, 0.91, 0.197, and 0.908, respectively. Fig. 4 illustrates the Gaussian plume dispersion model. Note that although the Gaussian plume model might not be applicable (due to turbulence) during high (> 6 m/s) wind velocities, such wind speeds are considered uncommon, and sensor readings during those times will be ignored without significantly compromising the desired 18 day detection time. Similarly, the readings at wind speeds below 1 m/s will be omitted.

If valid, the Gaussian plume distribution at a point $x$ downstream of the leak is uniquely defined by the dispersion parameters, which are assumed constant, the wind velocity $U$ (m/s) and the emission rate at the leak $Q$.

\section*{B. Single Plume Leak Estimation}

Once a plume moves across a sensor location due to variations in wind direction, a plume distribution can be reconstructed. This concept is illustrated on Fig. 5. As CH\textsubscript{4} plume from a nearby leak rotates from position (i) to position (ii), the sensor signal records the plume profile. The below analysis assumes the wind velocity is constant, which may not be the case; however, the theory extends easily to cases with variable wind velocity.

Calculated distributions of a CH\textsubscript{4} plume using Eq. (1) are plotted on Fig. 6 as a function of rotating wind direction for two different wind velocities (1 ms\textsuperscript{-1} and 5 ms\textsuperscript{-1}). The distribution of methane concentration was calculated at different points along the centerline of the plume (2 m – 14 m distances from the leak location). The mass emission rate
was assumed to be 0.032 g/s or 6 standard cubic feet per hour (SCFH).

The down-wind coordinate \( x \) from the leak (the distance to the sensor from the leak along the centerline of the plume), can be extrapolated from the peak concentration \( \lambda \) (ppm) and wind velocity \( U \) (m/s) by numerically solving the following equation:

\[
\log( ca ) + (b + d) \log x = \log \left( \frac{Q V_m}{M \pi U} \right) - \frac{(\Delta h)^2}{2a^2x^2b} \tag{4}
\]

where, \( \lambda \) is the concentration of the detected \( \text{CH}_4 \) in parts per millions (ppm), \( M \) is the Molar mass of \( \text{CH}_4 \) (g/mol), \( V_m \) is the molar volume of an ideal gas at standard temperature and pressure. The plume rise due to the buoyancy of \( \text{CH}_4 \) (\( \Delta h \)) is calculated as

\[
\Delta h = \frac{E x^\beta}{u^\alpha} \tag{5}
\]

where, for a stable atmosphere, \( \alpha = 1 \), \( \beta = \frac{2}{3} \), and \( E = 1.6 F^{(1/3)} \), where \( F \) is the buoyancy flux parameter, defined as

\[
F = \frac{gd^2V_s(T_s - T_a)}{4T_s} \tag{6}
\]

where \( g \) is acceleration due to gravity (9.8 m/s\(^2\)), \( d \) (m) is stack diameter, \( V_s \) (m/s) is stack exit velocity, \( T_a \) (°C) is ambient temperature at stack height, and \( T_s \) (°C) is stack exit temperature at stack height.

Once distance to the leak \( x \) is established, the following formula can be used to calculate the leakage rate:

\[
Q = \frac{\lambda M}{\pi \sigma y \sigma z U} \times \exp\left[ \frac{1}{2}\left(\frac{y}{\sigma y}\right)^2\right]
\times \exp\left[ \frac{1}{2}\left(\frac{H}{\sigma z}\right)^2\right] \tag{7}
\]

where \( Q \) (g/s) is the mass emission rate (MER) of the leak. Consequently, estimation of the MER of a single leak is possible by scanning the leak distribution plume.

The mass emission rate (MER) of the leak (g/s) as a function of maximum detected \( \text{CH}_4 \) concentration by sensors located at various distances along the plume centerline is illustrated in Fig. 7.
The emission rate was assumed to be 0.032 g/s or 6 SCFH. The wind speed is assumed to be 5 m/s. The mass centerline of the plume (2 m – 14 m distances from the leak position). The wind speed is assumed to be 5 m/s. The mass emission rate was assumed to be 0.032 g/s or 6 SCFH.

The peak concentration of a CH₄ plume at its centerline as a function of wind speed (1 m/s – 5 m/s) is plotted in Fig. 9. These results are based on a simulation that assumes the leak is at ground level and that the mass emission rate from the leak was assumed to be 0.032 g/s or 6 SCFH and held constant. The distribution of methane concentration was calculated at different distances from the methane leak (2 m–14 m away from the leak). The plume rise due to the buoyancy of CH₄ (Δh) is included in the plume distribution calculation. The response of the sensor is larger at slower wind velocities. As the wind velocity increases the sensor response becomes smaller following a simple logarithmic decay (Fig. 9). Our calculation also suggests that considering an LDL of 2 ppm, the sensor will be able to detect CH₄ leaks in the range of 0.09 m/s to 9 m/s of wind speed provided that the sensor is located at 2 m distance on the centerline of the plume.

C. Quantifying Sensor Response and Noise

The shape of the plume recorded by the sensor depends on its lower detection limit (LDL). Single ppm CH₄ sensitivity is currently achieved using our MONC-MWCNT chemoresistor sensor [16], [17]. Consequently, we defined the LDL of the sensors as 1 ppm. Our ongoing research is focused on achieving sub-pmm detection limit. For this particular analysis, we estimated the integration time of the sensor as one minute. Assuming an error of ±LDL, the error in plume height and plume width can be estimated. Errors in other factors, such as wind speed or wind direction will also contribute to the error of the sensor, but can be reduced by increasing integration time.

Fig. 10 shows the center point plume concentration as a function of wind speed (1 – 5 ms⁻¹) and distance, for a point leak with mass emission rate of 6 SCFH. The error bars (Fig. 10) indicate the distance uncertainty at 8 m distance from the leak (Note 14 m is the largest distance from the leak in a 10 m × 10 m grid).

The wind data (National Wind Technology Center dataset obtained near Boulder, Colorado, US [18]) suggests that errors contributed by wind velocity and wind direction can be reduced by improving the data integration time of the sensor. To obtain better understanding on the errors related to wind speed/direction and the integration time of the sensor, future work should involve further simulations based on real-time measured wind velocity and wind direction, i.e., to model variation in wind direction and velocity in different time segments and error estimation based on sensors response time.

V. UV-RECOVERY-BASED PLUME MAPPING ALGORITHM

Once a chemoresistor CH₄ sensor is exposed to CH₄, the response back to baseline resistance, i.e. recovery, is slow [13], [16]. We have developed a UV recovery technique for the ZnO-MWCNT CH₄ sensor that improves the recovery time (about 3 min) by a factor of 10 compared to published studies (about 30 min) [13] as well as our previous work [16]. Optoelectronic application of the ZnO-CNT, such as optical switching has been reported before [20], [21]. ZnO is a wide-bandgap semiconductor (3.37 eV) with a large exciton binding energy (60 meV) [21], which facilitates photo-induced enhancement of sensor recovery. UV irradiation lowers the energy barrier of desorption of CH₄ molecules from the functionalized CNT surface [13], which increases the removal of CH₄ from the CNT surface and shortens the recovery time of the sensor [13]. Once the electron-hole pairs are generated in the functionalizing ZnO nanoparticles as a result of UV exposure, electrons migrate to the surface and help to desorb the CH₄ molecules [22].

Results presented in this section show that there is an improvement in the recovery time resulting from the UV exposure of the MONC-MWCNT sensor. To study the UV recovery, the sensor was first exposed to 10 ppm CH₄ in air for 30 min. Without interrupting the flow of CH₄, the sensor was irradiated with a UV light (390 nm) until the sensor returned to its baseline resistance (Fig. 11). A recovery time of about 3 minutes was observed, which is a 10 times improvement compared to past results even when the CH₄ flow was interrupted [13], [16].

Consequently, a UV-recovery-based mapping algorithm has been proposed for the ZnO-MWCNT sensor. The flowchart of the UV-recovery plume mapping algorithm is shown in Fig. 12. The complete mapping mechanism is decomposed into three phases. In the first phase, the sensor is initiated to sense the plume (phase 1: initiate search for plume). In phase 2, when a plume hits the sensor, it starts to respond. The value will be set little higher than the maximum resistance to avoid false detection of the plume. A resistance increment of 11%
Fig. 11. Relative resistance ($\Delta R/R = (R_{CH4} - R_{air})/R_{air}$) of the ZnO-MWCNT sensor while exposed to 10 ppm CH$_4$ in air and recovered by irradiating with a 390 nm wavelength UV light (without stopping the CH$_4$ flow). Average recovery time was about 3 min. After each UV exposure, the sensor peaked in about 15 min.

(defined from the result presented in Fig. 11), i.e., when the sensor’s resistance crosses the maximum value, the next phase is triggered and the sensor sends output data. Simultaneously, the UV is turned ON (phase 3: send output data; UV ON). In the next phase the sensor response is compared with baseline value. UV will be off if the sensor reaches its baseline; otherwise, further recovery will be carried out. The UV-based recovery reduces the recovery time and consequently detection time of MONC functionalized MWCNT methane sensors by utilizing strong UV response of functionalizing MONCs.

VI. SENSOR RESPONSE TIME AND INTERFERENCES

The current generation of the sensors showed response time of 15 min (Fig. 11) It has been reported that the long response time of CNT chemoresistive sensors might be a result of the slow diffusion and weaker absorption of the gas molecules onto the sensor surface [23]–[26].

Reduced dimension of the functionalizing ZnO nanocrystals might help improve the diffusion and absorption of the gas molecules by creating a larger number of active sites, i.e., more metal oxide sites. Reducing the dimension of the ZnO nanocrystals can be achieved in two ways: (1) reducing the number of ALD cycles when functionalizing the MWCNTs with ZnO, which reduces the thickness of the ZnO NCs and improves their discreteness on the MWCNT surface and (2) in related work, we observed that the mean size of the functionalizing ZnO NCs on MWCNTs becomes smaller as the ALD temperature increases [16]. The crystalline quality of the functionalizing ZnO NPs also improves with a higher temperature ALD. In the future, we plan to fabricate sensors with smaller sized NPs by increasing the ALD temperature. We expect a shorter response time from this next generation of sensors.

It was reported that ZnO functionalized CNT based chemoresistor sensors are sensitive to CO [27], NO$_2$ [28], H$_2$O, O$_2$ and H$_2$ [29]. To study the response of the sensor to relative humidity (RH), the resistance of the ZnO-MWCNT sensors at different RH was measured at room temperature by controlled flow of moist air inside an enclosed plastic chamber. The resistance of the sensor increased monotonically as the RH was increased from 10% to 91%. The resistance of the sensor returned to its initial value as the RH was reduced back to 10%. The highest recorded relative resistance change ($\Delta R/R = (R_{water} - R_{air})/R_{air}$) or the ZnO-MWCNT sensor was about 4% at 91% RH. Several preliminary tests with CO$_2$ were also performed. The ZnO-MWCNT sensor showed about 1.5% relative resistance at room temperature while exposed with 650 ppm of CO$_2$. However further experiments are required to verify sensitivity. Future plans include exposing the sensor to relevant levels of NO$_2$, H$_2$, and O$_2$ as well as additional testing with CO$_2$.

To address interference from other hydrocarbons present in natural gas, primarily ethane, we anticipate to use a combined signal from multiple sensors functionalized with different catalysts, akin to the electronic nose concept [30].

VII. CONCLUSION

We have fabricated a metal oxide nanocrystals (MONC) functionalized MWCNT-based CH$_4$ chemoresistor sensor capable of sensing below 10 ppm of CH$_4$ in dry air at room temperature. We presented a Gaussian plume model (GPM) based algorithm for a distributed methane sensor system (DMSS) which can use our chemoresistor sensor to localize leaks in natural gas infrastructure, such as production wells and pads. We present a localization algorithm, which can be used with the distributed sensors to pinpoint CH$_4$ leaks, allowing the DMSS to be used for continuous low-cost monitoring of CH$_4$ emission. We were able to reduce the recovery and consequently detection time of
functionalized MWCNT-based CH4 sensors by utilizing strong UV response of functionalizing MONCs. Finally a UV-recovery-based plume detection algorithm is presented which offers acceleration in overall plume mapping processes. Together with energy harvesting and ultra-low power wireless solutions this system can be used for ubiquitous monitoring of natural gas infrastructures such as production pads, valve stations and transmission pipelines.

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REFERENCES


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