Environmental Monitoring for Smart Cities

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Abstract—This work presents an innovative, multidisciplinary and cost-effective ecosystem of ICT solutions able to collect, process and distribute georeferenced information about the influence of pollution and micro-climatic conditions on the quality of life in Smart Cities. The system has been developed and experimentally evaluated in the framework of the research project SHE, co-funded by the Tuscany Region (Italy). Specifically, an innovative monitoring network has been developed, constituted by fixed and mobile sensor nodes, which provided comparable measurements in stationary and mobile conditions. In addition, sensor data have been enriched with those generated by citizens through the use of a dedicated mobile application, exploiting participatory sensing and MSN paradigms.

Index Terms—environmental monitoring, mobile sensor networks, smart city, participatory sensing, mobile social networks

I. INTRODUCTION

The idea of Smart Cities (SCs) has its roots in the so-called Healthy Cities programme [1], launched in 1987 by the World Health Organization (WHO) and still active. It is a long-term international development initiative aimed at placing citizens’ health in urban areas high on the agenda of the decision makers, to promote comprehensive local strategies for health protection and sustainable development. Currently, WHO is focusing on the so-called Health 2020 programme (phase VI), exploiting whole-of-government and whole-of-society approaches. The latter approach explicitly takes into account the views of citizens, by considering and acknowledging the relationship between people and governors, in order to deal with needs and issues as soon as they arise. In the framework of SCs, the air quality is one of the fundamental factors that should be constantly monitored because of its effects on health [2], [3], [4] and, more generally, on citizens’ Quality of Life (QoL). In fact, WHO launched the Global Platform on Air Quality and Health, calling for a collaborative effort in order to develop, implement and monitor air pollution abatement strategies. In this context, the Smart Healthy Environment (SHE) project designed and deployed an ecosystem of Information and Communication Technology (ICT) solutions aimed at monitoring the environmental conditions and actively including citizens in the generation and exploitation of useful information. The first objective was to develop a cost-effective, distributed and efficient sensor network for collecting, processing and distributing data related to the air quality in the city of Pisa, Italy. It consists of both fixed and mobile sensor nodes able to measure several environmental parameters. Through the use of the mobile nodes, we investigated the feasibility of a totally mobile sensor network for urban environmental monitoring and the impact of the mobility of the nodes on the measurement accuracy. Mobile sensor nodes offer several advantages: (i) they can extend the coverage of the environmental monitoring at a very low cost, thus complementing and/or substituting data derived from fixed nodes; (ii) a high spatial and temporal resolution is achievable, providing a statistically reliable measurements flow; (iii) the maintenance of the nodes is simplified, not requiring interventions of specialised technicians in potentially distant locations; (iv) the malfunctioning of a single sensor node can be simply identified by comparing its measurements with those provided by close devices. In order to evaluate whether or not the mobility of nodes can affect the quality of the measurements, a real test-bed has been conducted in the city of Pisa by comparing the measurements collected by fixed and mobile sensor nodes built within the SHE project.

As a second goal, the sensor network extends the environmental monitoring by exploiting the personal mobile devices of the citizens, allowing them to report and comment on those situations that can further influence the environmental conditions and, more in general, QoL in the city. Specifically, a Mobile Social Network (MSN) application, namely SmartCitizen, has been developed to implement the participatory sensing paradigm, by directly collecting and sharing contents provided by citizens through their personal mobile devices (i.e., smartphones and tablets). Citizens provide subjective information (e.g., comments, pictures, videos) that represents an additional support to the objective information collected by the sensor nodes. In fact, user-generated contents are both distributed among citizens through the opportunistic communication of their mobile devices, and stored on the central host of the herein proposed architecture, in order to be correlated with data collected by nodes. Then, this information is made available to citizens, city governors, local authorities, and all the other experts interested in receiving timely updates on the air quality in the city.

In order to make the information accessible to citizens in a simple way, we defined three indexes derived from sensor data: Air Quality Index (AQI), Thermal Comfort Index (TCI), and Traffic Index (TI). The last one is obtained by an additional sensor network already present in the city of Pisa and managed by a local company, Pisa Mobility (PisaMO). Accessing the
In the last few years several solutions for environmental monitoring has been presented in literature. The authors in [5] propose an air pollution analysis based on meteorological and traffic data collected in Milan during 2013. The following parameters were collected: (i) pollutant concentrations, including Particulate Matter (PM) 10 and 2.5, carbon monoxide (CO), ozone (O₃), nitrogen dioxide (NO₂), benzene (C₆H₆), and sulfur dioxide (SO₂); (ii) climate conditions, such as air temperature, relative humidity, precipitation level, wind speed and atmospheric pressure; (iii) traffic data for different categories of vehicles. Monitoring stations were deployed along the city; meteorological data were gathered from a simple personal weather station and, finally, traffic data were collected thanks to the use of Wireless Sensor Networks (WSNs). A web-based interface was developed to provide access to the data and the results of the data analyses. However, the whole system in [5] is based on the use of fixed installations only. The use of mobile sensor nodes has been recently studied in [6], where sensors are placed on rentable bicycles. Every time a bicycle is in use, the sensor collects data and stores them on the available local storage; when the bicycle is returned, the collected data, geo-referenced thanks to a Global Positioning System (GPS) receiver, are sent to a central server via a General Packet Radio Service (GPRS) connection. The authors aimed at demonstrating that the measurements collected by the mobile sensor nodes do not depend on the sensor orientation and on the sampling period; instead, the actual path is crucial to correctly evaluate the results, especially to estimate the distribution of pollutants in a target area. However, the authors could not effectively demonstrate the accuracy of the mobile measurements as the samples collected by the mobile sensor nodes were compared to indoor measurements, hardly comparable with outdoor ones. Crowdsourced air quality monitoring is studied in [7], where also the physical activity of volunteers has been recorded, in order to assist environmental health studies. Low-cost mobile sensor nodes have been used during a campaign that lasted half a year in the city of Gjøvik, Norway, proving the feasibility of a low-cost crowd-sourced data collection platform. Anyway, the authors only proved the feasibility of such a platform, since the campaign had a limited duration. In [8], the authors propose an air quality monitoring system, namely uSense, based on the use of small and low-cost nomadic devices, running on batteries. Those sensor nodes can be moved from one place to another when the need for readings in a new area arises: only O₃, NO₂, and CO are monitored and transferred via a wireless connection to a central server. The server estimates an air quality index, accessible to users via a web interface or a mobile application.

The main difference between our solution and those previously cited consists in the realization of a complete ecosystem for environmental monitoring, relying on a distributed WSN, composed of both fixed and mobile nodes, and on the participatory sensing paradigm, by directly involving citizens through their personal mobile devices, according to a whole-of-society approach. The ecosystem could also be extended with wearable devices aimed at further increasing the coverage area of the monitoring system and also at providing a different perspective in the collection of environmental data, e.g., to analyse the individual and collective exposure to pollutants. A first attempt to use these type of body-worn sensors has been proposed in [9] by using a prototype sensor board. Currently, some emerging companies are investing in the development of such devices trying to guarantee a comparable level of accuracy with respect to certified sensor stations. However, they are still not available in commerce at this moment and, for this reason, we decided to initially focus on the explicit user contribution to enrich sensor data as fundamental part of participatory sensing paradigm. In addition, SHE ecosystem is open to the integration of heterogeneous devices and external monitoring systems implementing the OGC SWE standard for sensor data encoding. To this aim, SmartCitizen app supports the same standard directly on mobile devices by integrating an apposite software framework to optimise sensor data management with respect to the limited resources of mobile devices. SHE data is also accessible in customised ways, according to the users’ categories (i.e., common citizens, experts, local authorities), both through the mobile app and the project website, in order to ensure a global comprehension of the system’s results.

III. Scenario and System Architecture

The system architecture we designed is depicted in Figure 1. The objective (derived from sensors) and subjective (generated by citizens) data are collected from heterogeneous nodes in order to derive useful indicators and to generate a space for discussion among citizens and between citizens and local governors. By only relying on the objective data, AQI, TCI and TI indexes are computed and subsequently presented to different user categories: citizens, scientists, and local authorities.
Fig. 1: A logical description of the SHE architecture: objective and subjective data are collected from fixed and mobile sensor nodes, in order to derive air quality indicators, presented to different types of users via website and mobile application. The citizens are the core of the system, representing both data producers and consumers.

![Fixed Weather Station](image1)

![Sensor Node on Bicycle](image2)

Fig. 2. The sensor node and the weather station in use in the proposed architecture for monitoring the air quality.

The objective data are then complemented with the subjective data generated by citizens through the SmartCitizen MSN application. Our sensor platform collects the following environmental parameters: CO, carbon dioxide (CO₂), unburned hydrocarbons (HC), O₃ and PM 2.5 pollutants, in addition to micro-meteorological parameters. n = 9 sensor nodes have been deployed in the city of Pisa: seven are fixed ones, two are mobile ones; furthermore, we also used a weather station along with a solar panel to recharge internal batteries (see Figure 2a). The sensor node we developed has been used as fixed and mobile one, and is visible in Figure 2b, installed on a bicycle. The fixed sensor nodes have been placed on two intersecting paths in the city center: (i) a heavy traffic path, thus expecting higher pollutants concentrations, (ii) and a fitness path, a low-traffic path close to the city center and used by citizens for outdoor fitness activity. A sensor node used as network coordinator has been placed at the intersection point of the two paths. The distance between two consecutive fixed nodes is approximately 350 meters on the heavy traffic path and approximately one kilometer on the fitness path, so that the whole network covers an area of approximately 4 to 5 square kilometers, corresponding to Pisa’s city center. Figure 2 shows the position of the fixed, mobile and traffic sensor nodes, as visible on the website. Several colored icons are shown per physical node (refer to Section III-A). The traffic sensors are placed at the entrance of the main arterial roads from and towards the city center. The instantaneous position of the mobile sensor nodes is visible in the top center of Figure 2; they are identified by a white icon surrounding a cycle. The base sensor node is adaptable and easily expandable: it is provided with large processing capacity, a Wireless M-Bus (WMBus) radio in the case of fixed nodes, WMBus and ZigBee radios in the case of mobile nodes, a solar cell to recharge the internal battery, and programmable analogue and digital components to interface with the on-board sensors. WMBus has been used for the data exchange among the fixed nodes. ZigBee has joined the WMBus protocol in the case of mobile nodes, aiming at studying the operative limits of the ZigBee protocol when used in vehicular contexts at urban speeds. The mobile nodes have been installed on bicycles traveling around the city (but cars and buses can be used, too), as for instance visible in Figure 2b. We also performed an additional study by using an Unmanned Aerial Vehicle (UAV) carrying a simplified sensor node as payload, in order to study the quality of the transmission channel. The results of the latter study are reported in [13], [14]. The communication infrastructure allows both the communication among the nodes and the communication between the coordinator and the central host (acting as server), where the collected data are stored and processed. The central server is hosted at a remote location, and it provides a database service for storing collected data, analysis procedures, control interfaces for experts and technicians, and a public open interface accessible via the aforementioned website.

A. Sensors and communication platform

We identified the following sensors for the detection of the aforementioned environmental parameters:

- Sensirion SHT75 (temperature, relative humidity and dew point);
- Non-Dispersive Infrared Sensor (NDIR) gas sensor (CO, CO₂, HC), namely G3, based on Dual Wavelength Ratioing (DWR) technology;
- SGX Sensortech MiCS-2614 (O₃);
- Davis Instruments 7911 and WS1070/WS (anemometers);
- Davis Instruments 7859 (rain gauge);
- Arcus RPTF 2 PT1000 (radiant temperature);
- Qbit MP25 (PM 2.5);
• Davis Instruments 6490 (radiometer);
• Freescale MPL3115A2 (atmospheric pressure);
• GPS SoC Mediatek MT3329 (mobile nodes only).

The specification of the aforementioned commercial sensors (apart from the G3 one) can be found online. G3 sensor has been specifically designed and built within the project activities, because no commercial product was found, at that time, satisfying the system requirements. Electrochemical sensors have been excluded because of their limited average lifespan, in favor of NDIR technology. The G3 sensor performs four IR measurements in four different optic bands, and the response time (T90) is of the order of tens of seconds (< 1.5 minutes).

In order to have accurate measurements, the gas sensor must reach the optimal exercise temperature, and it may take up to 15 minutes before the first reading is available. The limit of detections (LoDs) are: < 5ppm for CO, < 10ppm for CO2, and < 5ppm for HC, with a resolution of 1ppm for CO, 1ppm for CO2, and 1ppm for HC. The node is powered with a lithium battery operating at 3.7 V. Qbit Optronics developed the gas sensor according to the provided specifications, while the control electronics of the gas sensor have been designed and realized by project partners. All sensor nodes have been calibrated in laboratory before use and have been re-calibrated every three months, on average. The G3 sensor node built within the project has been calibrated (one point calibration at ≈ 70% of the span value) by relying on calibration gas cylinders.

To summarize, three types of nodes have been used:
• a weather station node: measuring temperature, relative humidity, dew point, wind speed and gusts, PM 2.5, HC, CO2, CO, O3, gas temperature, gas relative humidity, gas atmospheric pressure, radiant temperature.
• two mobile nodes: measuring temperature, relative humidity, dew point, HC, CO2, CO, O3, gas temperature, gas relative humidity, gas atmospheric pressure.

Each fixed or mobile node measures the aforementioned pollutants every $T_N = 5$ minutes and immediately transmits the results to the coordinator node, which transmits the collected measurements every $T_I = 15$ minutes to the central server via a GPRS connection. Therefore, $nT_I/T_N$ reports are received by the central server every $T_I$ minutes, triggering the update of the aforementioned indexes. The fixed nodes rely on the WMBus protocol (operating at 169Mhz) to exchange data with the coordinator, while the mobile nodes rely on the WMBUS/ZigBee protocol (the latter operating at 2.4 GHz) if in the communication range of another sensor node or, alternatively, on a GPRS connection.

B. Air Quality, Thermal Comfort and Traffic indexes

As already mentioned, we defined a set of indexes aimed at providing some clear indications on the city’s environmental conditions.

1) AQI: information on the air quality is conveyed to the citizens thanks to two sub-indexes: the first one, namely $AQI_{O3}$, is based on the ozone measurements; the second one, namely $AQI_{PM}$, is based on PM 2.5 measurements. $AQI_{O3}$, according to WHO guidelines, should be less than a mean
concentration of 100 µg/m³ within an 8-hour time window; AQI<sub>PM</sub>, according to WHO guidelines, should be less than a yearly mean concentration of 10 µg/m³. The sub-indexes are updated by applying a moving average every 8 hours in the case of O₃, and every 24 hours in the case of PM 2.5. Table I provides details on the thresholds in use for evaluating AQI sub-indexes, as shown in Fig. 2 (AQI<sub>O₃</sub> is marked as <i>o₃</i> and AQI<sub>PM</sub> is marked as <i>p₃</i> in Fig. 2). The value of the indexes does not provide any forecast.

2) TCI: it is based on the Universal Thermal Comfort Index (UTCI), which is a readily accessible thermal index based on a state-of-the-art thermo-physiological model [13]. Several applications and services are based on the use of such an index, for instance Public Weather Services and Public Health Systems, as well as precautionary planning and climate impact research in the health sector [15]. TCI value is updated every 15 minutes, using as an input the air temperature, the mean radiant temperature, the wind speed, and the relative humidity. The index is presented in the same color-coded way of AQI index; in Figure 2 it is marked as °C, and its thresholds can be read in Table II. Also in this case, the value of the index does not provide any forecast.

3) TI: the index in (1) estimates the traffic in cities assuming that a continuous flow of vehicles passes a virtual line, in the following referred to as access, where the traffic is measured. The index has the following formulation:

\[ TI = s_b K_1 K_2 K_3 K_4 \text{ [EV/s]} \]  

where \( s_b \approx 1800 \) is a constant value, representing a base congestion factor, while \( K_i \) factors are used to adjust the latter to the actual situation. The index is measured in Equivalent Vehicles (EV) per second. The most common vehicle classes (or types) are reported in Table IIIa, according to the Italian regulations. More specifically, \( K_1 = \frac{1}{\sum_i a_i E_i} \), \( \sum_i a_i = 1 \) describes the composition of the traffic (cars, motorcycles, trucks), as reported in Table IIIa; \( K_2 = 1 \pm 0.03 i \) takes into account the steepness \( i \% \) of the access: it decreases in case of uphill and increases in case of downhill; \( K_3 \) considers the specific position of the access in the urban area (for instance, centre or suburbs) and Table IIIb reports its possible values; \( K_4 = \sum_i b_i G_i \) considers the interference due to the presence of pedestrians and to the maneuvering of vehicles, as reported in Table IIIc. Twenty-two points for measuring traffic flows are placed around the city of Pisa, tracking vehicles entering and leaving the city, as well as the class of the vehicles.

### C. SmartCitizen MSN application

The SmartCitizen app aims at stimulating the active participation of citizens in collecting and sharing useful data related to QoL in their city, not only on the environmental conditions, but also on concurrent events or experiences that can provide additional information on the environmental situation. The app intuitively visualizes data provided by the deployed WSN, which provides the indexes discussed in Section II-B in a general and simple presentation consisting of colored circles, centered on the current geographical position of the sensor nodes. The app also stimulates discussions among citizens on several topics exploiting features similar to social networking applications, such as posting, commenting and chatting. The main difference between SmartCitizen and standard Online Social Networks is that citizens share their contents and experiences via Device to Device (D2D) communications, relying on proximity and opportunistic communications (as the main principles of MSN applications [16]) and avoiding a continuous storage on a centralised infrastructure. Figure 5 shows some screenshots of the graphical user interface of the app. In the first screenshot, the app presents the air quality index on a map, as measured on each sensing station deployed in the city, and presented as colored circles. The app is also designed for expert users able to interpret the detailed information about the measured quantities. In this case, the authorized users can visualize the detailed data of a station simply by clicking on the circle. All sensing data are downloaded from the central server, if the user device has Internet connectivity, or they can be downloaded through D2D communication, if available on other users’ devices in close proximity. The application also provides the user with the possibility of creating new posts, in which he can open discussion on a specific topic, tagging it for an optimised dissemination among users and devices. Finally, the user can visualise the list of active discussions on different topics, and the list of generated contents for each discussion with appropriate notification of new available contents. We also integrated additional features designed for sports users, considering the increasing trend on outdoor fitness activities, the lack of useful information about running paths conditions and, overall, on the healthy conditions of specific city areas. Specifically, SmartCitizen allows users in close proximity to share fitness experiences in a quasi real-time scenario. In this case, they can directly record activity paths in the city, and enrich the data with their own information, comments, and suggestions. The air quality information in the area is automatically disseminated among the users’ devices through a context-aware content dissemination protocol provided by
### Vehicle classes

| Vehicle | E  |
|---------|----|
| bicycles | 0.2 |
| motorcycles | 0.33 |
| cars | 1 |
| trucks | 1.75 |
| buses | 2.25 |
| trams | 2.5 |

### Localization

| Parameter | Value |
|-----------|-------|
| residential | 1 |
| commercial | 0.98 |
| industrial | 0.93 |
| business | 0.85 |

### Maneuvering types

| Type | Value |
|------|-------|
| straight line | 1 |
| turning right | 1 - 1.25 |
| turning left | 1 - 1.75 |

### TABLE III: Values of the parameters used for the estimation of the Traffic Index (TI)

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#### Fig. 3: Some screenshots of the MSN application, namely *SmartCitizen*, developed inside the SHE project

The deployment of the described WSN allowed us to conduct an extensive experimental evaluation of the feasibility of using mobile sensor nodes (instead of fixed stations) for environmental monitoring in Pisa, in addition to the evaluation of the air quality in the city. Mobile nodes are carried by a mobile entity (humans included), thus this type of network can provide accurate measurements on the individual exposure to single pollutants, further to environmental information on city areas. In the following, we focus on the comparison of data provided by (i) fixed nodes on the heavy traffic and the fitness paths, and by (ii) mobile and fixed nodes. The results of the statistical comparison are based on a measurement campaign in spring 2015 in Pisa. It is worth recalling that mobile and fixed nodes share the same hardware and software components for measuring the aforementioned pollutants, thus the overall setup is tested against mobility conditions.

### IV. EXPERIMENTAL RESULTS

The deployment of the described WSN allowed us to conduct an extensive experimental evaluation of the feasibility of using mobile sensor nodes (instead of fixed stations) for environmental monitoring in Pisa, in addition to the evaluation of the air quality in the city. Mobile nodes are carried by a mobile entity (humans included), thus this type of network can provide accurate measurements on the individual exposure to single pollutants, further to environmental information on city areas. In the following, we focus on the comparison of data provided by (i) fixed nodes on the heavy traffic and the fitness paths, and by (ii) mobile and fixed nodes. The results of the statistical comparison are based on a measurement campaign in spring 2015 in Pisa. It is worth recalling that mobile and fixed nodes share the same hardware and software components for measuring the aforementioned pollutants, thus the overall setup is tested against mobility conditions.

#### TABLE IV: Comparison between the average values collected on the heavy traffic ($M_{tr}$) and on the fitness ($M_{fit}$) paths.

| Measured quantity | $M_{tr}$ | $M_{fit}$ | $\eta_{TF}$ |
|-------------------|----------|----------|-------------|
| wind speed        | 0.62 m/s | 0.69 m/s | 0.12        |
| temperature       | 16.2 °C  | 14.7 °C  | 0.09        |
| relative humidity | 66.4%    | 70.2%    | 0.037       |
| dew point         | 9.7 °C   | 9.8 °C   | 1.04        |
| radiant temperature| 16.8 °C | 15.1 °C | 0.04        |
| PM 2.5           | 17.7 µg/m³ | 14.8 µg/m³ | 0.16        |
| unburned hydrocarbons (HC) | 3.12 ppmV | 3.12 ppmV | 0.0006 |
| carbon dioxide (CO₂) | 423.26 ppmV | 451.1 ppmV | 0.06 |
| carbon monoxide (CO) | 2.05 mg/m³ | 2.28 mg/m³ | 0.11 |
| ozone (O₃)        | 48.91 µg/m³ | 51.33 µg/m³ | 0.05 |

Figure 4 shows the comparison between Probability Mass Functions (PMFs) of the considered pollutants and quantities collected by the sensor nodes located on the heavy traffic path and on the fitness path. The average values of PMFs in Figure 4 are reported in Table IV which also compares the readings by relying on the relative error definition: $\eta_{TF} = \left| 1 - \frac{M_{tr}}{M_{fit}} \right|$. The PMFs of the dew point and of HC are not visible in Figure 4 for space reasons ($\eta_{TF}$ is negligible for both quantities). Heavy traffic and fitness paths show similar average values because Pisa’s centre is quite small.

Figure 5 shows the comparison among PMFs of the pollutants and quantities collected by the mobile nodes with...
those collected by the closest fixed node, the numerical comparison (average values) is reported in Table V which also provides \( \eta_{MF} = 1 - \frac{M_{mob}}{M_{fix}} \). For the same reasons as before, the PMF of the dew point is not shown in Figure 5. In this case, the most significant differences are related to HC and O\(_3\). It is worth highlighting here that the fixed nodes collected data at a different height w.r.t. the mobile nodes (closer to the exhaust pipes of cars), thus part of the aforementioned skew should be ascribed to that. In addition, the skew depends also on the fact that the reading of mobile nodes are collected in an area close to fixed nodes (≈500m radius) and not in the same exact spatial position, thus introducing a further measurement error due to the mobility and to the coarse-grained spatial precision.

Since the large majority of the samples collected by the mobile nodes is geographically located in close proximity of two fixed stations, we considered both of them for the comparison.

The results presented in Tables V and VI indicate that the use of mobile sensor nodes for environmental monitoring is feasible (if HC is excluded), and it provides a discrete level of accuracy, with the advantage to extend the covered area in an easy and low-cost way. Furthermore, if a larger number of mobile sensors were to be deployed, the availability of large data-sets in even small geographical areas would provide statistically reliable measurement flows.

V. CONCLUSIONS

From a technical point of view, we can conclude that the use of mobile sensor nodes can be quite effective in monitoring the air quality in cities, increasing the monitoring area with limited costs, w.r.t. to the use of only fixed nodes, at the cost of a tolerable measurement error due to mobility and other factors, such as the height at which the mobile sensor
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