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1 **ROUTE2: A CLOUD-BASED INFRASTRUCTURE FOR ASSISTED TRANSIT**

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1 ABSTRACT

2 We introduce RouteMe2, a cloud-based system that was designed to facilitate use of public
3 transit by those who, due to visual or cognitive impairment, or old age, have difficulties traveling
4 independently. RouteMe2 is comprised of a software infrastructure (including a cloud server, a
5 web application, and a mobile application) and a physical infrastructure for fine-grained
6 localization at bus stops or at train platforms. In addition, RouteMe2 uses beacons placed inside
7 bus vehicles and train cars, which allow for identification of an incoming vehicle. Travelers or
8 other authorized individuals (family members, caregivers) can register a trip using the web
9 application. The traveler may receive specific notifications, such as when he or she reaches a
10 desired bus stop or a specific waiting/boarding area within the stop, or when the desired bus
11 vehicle has arrived. Authorized individuals may also track the traveler's trip remotely using the
12 web application, and be notified in case of problems (e.g., if the traveler has taken the wrong
13 bus). A pilot implementation of RouteMe2 was completed at the UC Santa Cruz campus, with a
14 demonstration of the most critical functionalities of the system.

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19 *Keywords:* Public transit, Accessibility, Localization, Beacons

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1 INTRODUCTION

2 A large number of people rely on public transportation for their mobility needs, and yet have
3 difficulties using it. Anybody who has taken a bus ride in a city he or she is not familiar with
4 may have experienced the difficulty, uncertainty, and sometimes anxiety associated with
5 determining which bus to take, and at which stop to exit the bus. For someone with poor vision
6 or blindness, taking a bus on an unfamiliar route can become a serious challenge, potentially
7 leading to dangerous situations, such as finding oneself lost in an unknown neighborhood.
8 Likewise, a person with some level of cognitive impairment may become easily confused or feel
9 unsure about which platform he or she should stand on while waiting for a train, or how to
10 negotiate a transfer between a bus and a light rail ride. ADA-mandated acoustic systems that
11 announce the number of an upcoming bus, as well as each stop in a bus or a train, are of great
12 help, but often cannot be heard because of loud background noise, or may be missed if someone
13 was not paying attention. Indeed, some people need to hear the same information multiple times
14 to comprehend it, and hear it again if they become confused.

15 While some of these people have family members who can drive them to places, this is
16 not a possibility for those who live alone, or when their loved ones are busy with work or school.
17 An alternative is represented by Paratransit, an ADA-mandated service provided by transit
18 agencies that offers individual rides on accessible vehicles. Passengers need to register with the
19 transit agency to reserve a ride the day before the trip, and pay a nominal fee (usually of a few
20 dollars). The cost of a Paratransit ride to the agency is in fact much higher. In rural areas,
21 Paratransit is de facto the only alternative available to its users (1). In urban environments,
22 however, the Paratransit system is widely perceived as economically inefficient with respect to
23 fixed routes (2). Hence, technology that could support safe and comfortable use of public transit
24 by people who would normally rely on Paratransit, would not only increase the independence of
25 these people (who would not need to make advance reservation for a ride), but also increase
26 revenues for the transit agencies.

27 This paper describes the initial development of RouteMe2, a service testbed comprising
28 three different components: (i) An app running on the passenger's smartphone; (ii) A Bluetooth
29 Low Energy (BLE) beacons infrastructure that enables fine-grained localization and
30 identification of oncoming bus vehicles or trains; (iii) A cloud service that receives data from the
31 user's smartphone, along with static and live feeds from the transit agencies, and produces
32 directions, notifications and warnings that are transmitted to the user's smartphone. RouteMe2
33 allows a user or an authorized person (family member or caretaker) to register a trip, possibly
34 with multiple legs and using multiple agencies, in a way similar to familiar trip schedulers such
35 as Google Maps. Once the trip is started, users receive continuous, context-aware information,
36 notifications, and warnings, designed to ensure that the trip is executed correctly. Authorized
37 persons can monitor the passenger's progress throughout the route remotely. The system, through
38 a dynamic tracking algorithm that uses all available information, can also detect when an
39 unexpected situation has occurred – if the user entered a wrong bus, for example, or if he or she
40 is waiting for the train on a wrong platform. In this case, RouteMe2 can issue an alert message to
41 the user and to a remote authorized person, who can take the necessary provisions (e.g.,
42 re-planning the trip, calling the traveler on the phone).

43 RouteMe2 is still in its initial development phases, with a pilot deployment completed
44 and demonstrated at the UC Santa Cruz campus. We believe that, once fully developed and
45 widely deployed, the RouteMe2 service could be tremendously helpful for travelers with
46 perceptual or cognitive impairments. It would facilitate independent usage of public transit,
47 enabling safe and comfortable rides. Family members and other individuals concerned with the

1 safety of the travelers would benefit from the ability to remotely monitor the trip, and be alerted
2 when something goes wrong. Ultimately, we believe that RouteMe2 would reduce the need for
3 Paratransit, increase ridership on fixed route transit, and help transit agencies meet their
4 accessibility goals.

6 **PRIOR WORK**

7 Several prior projects have addressed the problem of providing information access to travelers
8 with sensorial or cognitive impairments. For example, the ABLE Transit system (3) was
9 designed to provide location-contingent data (extracted from GTFS feeds) to blind travelers.
10 Azenkot et al. (4) used a Braille display connected to a smartphone to provide access to arrival
11 times (using the OneBusAway system (5)) for blind or deaf-blind travelers. Bluetooth beacons
12 were used in the Accessible Bus System (6) to alert travelers upon arrival of a desired bus. Wi-Fi
13 routers were used in the Public Transit Assistant (PTA) system to support independent travel by
14 blind persons (7)(8). These routers were installed both at bus stops and inside bus vehicles.
15 In-stop routers allowed a user to receive information about the stop location, the bus lines
16 through that stop, and arrival times. In-vehicle routers enabled notification of arrival of a desired
17 vehicle, and, during a bus ride, notification of upcoming stops.

18 Other projects have addressed the needs of passengers with cognitive impairments
19 (9)(10)(11). For example, the Travel Assistance Device (12) relied on the GPS in the traveler's
20 smartphone while riding a bus to determine the bus location, and to inform the traveler that a
21 desired stop was approaching (a similar system was described in (13)).

22 The use of BLE beacons for positioning has received increasing interest over the past
23 few years (14)(15). Pilot projects using BLE beacons in transit hubs include installations at
24 selected Massachusetts Bay Transportation Authority (MBTA) rail stations (16), at selected bus
25 stops managed by the Santa Clara Valley Transportation Authority (VTA, 17), and at Gatwick
26 Airport (18). An open standard for digital wayfinding using technologies such as BLE beacons,
27 specifically designed for blind and low-vision traveler, is being developed by Wayfindr.net.

29 **ROUTE2: GENERAL STRUCTURE AND PILOT DEPLOYMENT**

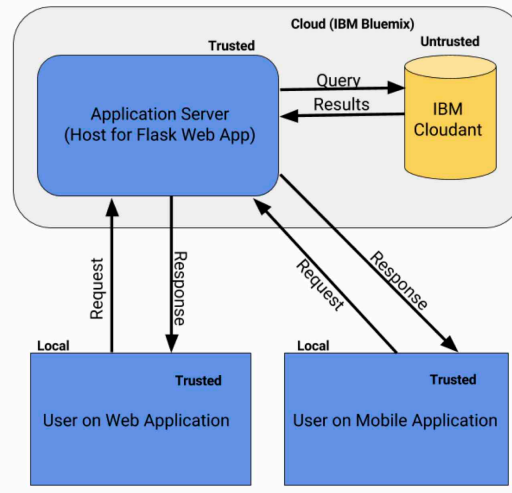
30 The RouteMe2 system is comprised of a software infrastructure (cloud database, web
31 application, and mobile application) as well as of localization infrastructure (BLE beacons
32 placed at bus stops and train platforms, as well as inside bus vehicles and train cars). In this
33 section, we provide details on both infrastructures, with a focus on the specific pilot
34 implementation that was completed at the UC Santa Cruz campus, demonstrating the main
35 functionalities of the system.

37 **Software Infrastructure**

38 The general software infrastructure of RouteMe2 is shown in Figure 1. Two applications are
39 available for users: a web application, which allows users to register a trip and to monitor the
40 trip's execution, and a mobile application, which runs on the user's smartphone and is used to
41 transmit location information to the cloud server and to receive notifications and directions. Note
42 that the web application can run on the user's smartphone, as well as on any computer or
43 smartphone connected to the internet. This allows authorized individuals to monitor the traveler's
44 trip remotely.

45 In order to support the desired functionalities of RouteMe2, we defined two different
46 user types: *travelers* and *supervisors*. A supervisor can plan and register a trip for the traveler,
47 and monitor the trip during its execution (and possibly receive alerts in special situations). While

1 a traveler can be his or her own supervisor, the ability to assign a supervisor role to a person
 2 other than the traveler could be useful in situations that require special assistance or supervision
 3 (e.g., when the traveler is a child, or a person being cared for by a family member or caregiver).



4
 5 **FIGURE 1 RouteMe2 software and cloud services infrastructure.**

6
 7 *Database*

8 Our implementation uses an IBM Cloudant NoSQL Document store as a database. The database,
 9 which is hosted in the IBM Bluemix cloud server, contains multiple types of documents
 10 generated by the application, including user documents (storing personal information), user
 11 relation documents (specifying a traveler's supervisor), trip template and trip instance documents
 12 (which list information such as start and end points and timing for a trip), and user location
 13 documents (which store the current and historical locations for the user, as well as other
 14 information such as the means of transportation).

15
 16 *API Design*

17 An API (Application Programming Interface) has been designed to enable creation and
 18 management of documents in the cloud server (e.g., updates of the traveler's location). The web
 19 app uses the Google Direction API to generate a route for a given trip, including transit and
 20 walking legs. A trip-specific *tolerance* value is used to determine when the traveler has strayed
 21 too far from the planned route, and needs to be re-routed. The application also estimates the
 22 current speed of the traveler while walking, based on the five most recent locations. This is used
 23 to estimate whether the traveler will be able to reach the next transit point in time to catch the
 24 bus as planned, or if re-routing (or determination of the next bus arrival time) is needed. In
 25 addition, the system determines whether the user has arrived and is waiting at a bus stop, on the
 26 basis of whether the location sent by the mobile application matches that of the next transit stop
 27 in the route. Note that this requires self-localization with higher precision than normally provided
 28 by GPS. As discussed in the next section, we use BLE beacons to achieve the desired localization
 29 accuracy.

30
 31 *Web Application*

32 The web application, also hosted by the IBM Bluemix cloud server, is written in Flask, a
 33 Python-based microframework (19). The front-end development within the web application uses
 34 a JavaScript framework known as VueJS (20). The web application allows one to log in, create

1 trip templates, plan and start trips, and view ongoing trips. Google's Direction API provides
2 familiar route visualization embedded within the generated web pages (Figure 2).

3
4
5 **FIGURE 2 Sample page created by the RouteMe2 web application displaying a planned**
6 **trip.**

7
8 The web application also displays the current user's location, as determined by GPS and/or BLE
9 beacons, along with the uncertainty radius (Figure 5), and information about the user status.

10
11 *Mobile Application*

12 We developed a simple prototype application, implemented on both iOS and Android platforms,
13 for the purpose of demonstrating the main functionalities of the RouteMe2 system. This
14 application supports transmission of positioning information from GPS and BLE beacons
15 (location updates are sent every 5 seconds), along with other fields such as traveler ID, trip ID,
16 and localization accuracy, among others. In addition, the mobile application is responsible for
17 detecting arrival of a bus as well as permanence of a traveler inside a bus, as determined by
18 connection with the in-vehicle beacon (as explained in the next section).

19
20 *Security*

21 Particular care has been taken to ensure that the information transmitted to the cloud server and
22 stored in the database cannot be accessed by unauthorized individuals. All sensitive user
23 information (including start/end points for a trip, location data, user credentials) is stored
24 encrypted in the database, with access only granted to the user that created the data and to his or
25 her designated supervisor. A lockbox encryption scheme (21,22) was implemented for this
26 purpose. Note that the Cloudant database never stores unencrypted data. Unencrypted data only
27 exists on the Web server, which can be designed to not use permanent storage (RAM-only). This
28 reduces the likelihood of compromise. Another security requirement was to ensure that a user
29 cannot become a supervisor for a traveler without the traveler's permission. This was ensured by
30 verifying (using SHA-256 hashes) that the data in the database documents that contain the
31 relation user-supervisor has not been modified.

32
33
34 **Infrastructure for Fine-Grained Localization**

35 Self-localization is critical for correct trip execution. For example, travelers need to know how to
36 reach a bus stop from their location. In addition, we would like to help a traveler identify a
37 specific waiting/boarding area at a bus stop or train platform (see e.g. Figure 3(a)). This would be
38 particularly important for travelers with low vision or blindness, who would otherwise run the
39 risk of standing too far from where the bus pulls over, potentially missing the bus if they cannot
40 reach the vehicle's door quickly enough.

41 Thanks to its almost universal availability (at least in the outdoors), GPS is the
42 localization system of choice for virtually all existing travel apps. Unfortunately, the accuracy of
43 GPS (which can be as low as tens of meters in urban situations), while adequate for applications
44 such as car navigation systems, is not sufficient for our purposes. For example, GPS typically
45 cannot differentiate between two bus stops across the street from each other, nor can it help one
46 locate the waiting/boarding area. In addition, GPS doesn't work underground (e.g. at a subway
47 station). For these reasons, we rely on an infrastructure of BLE beacons to improve the

1 localization accuracy when necessary.

2

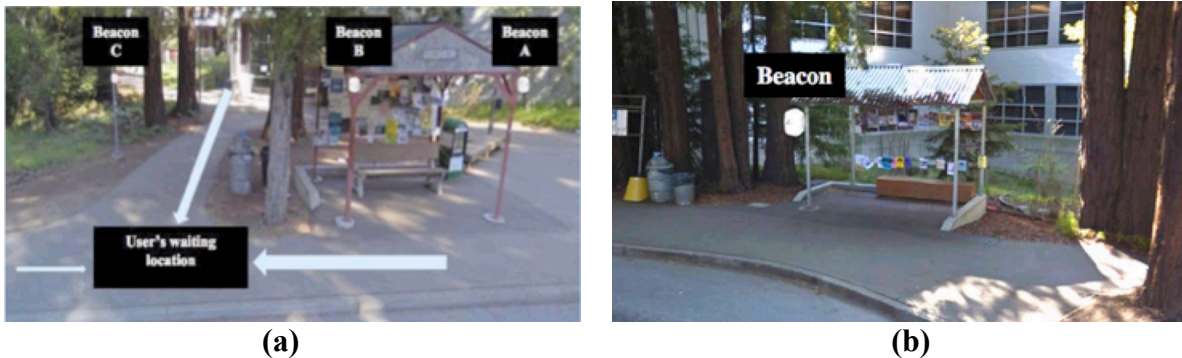


FIGURE 3 (a) The Science Hill – North bus stop, equipped with three Kontakt.io BLE beacons (shown magnified in the picture). The designated waiting/boarding area is also highlighted. (b) The Science Hill – South bus stop with a single beacon.

BLE Beacons

Over the past few years, Bluetooth Low Energy (BLE) beacons technology has been increasingly deployed for applications such as localization and proximity sensing. BLE beacons are inexpensive and typically run for months or even years on coin-cell batteries. There are two main BLE communication protocols (iBeacons and Eddystone), both supported by iOS and Android smartphones. For this project, we used Kontakt.io Tough Beacons, which, at the default power level, have a nominal range of 20 meters.

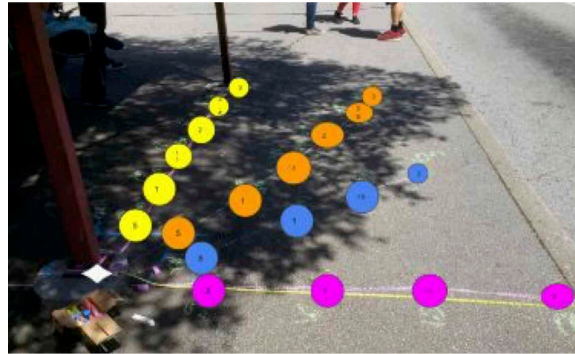
In our experiments, we instrumented two bus stops, one with a single BLE beacon, and one with three beacons (see Figure 3). If a single beacon is used, a proximity sensing modality can be enabled by setting a threshold on the received signal strength (RSSI), which is accessible via the Core Location Framework in iOS or the Proximity Beacon API in Android. By placing multiple beacons at different locations (but within transmission range from each other), it is possible to obtain relatively accurate localization from the set of RSSIs (*fingerprint*) from the beacons in range.

While it would be possible in principle to rely on multilateration using standard power decay (path loss) formulas (for the multiple beacons case), a more common practice is to learn the relation between RSSI fingerprints and location through an off-line calibration phase (sometimes called *wardriving*). Calibration requires building a database of RSSI fingerprints from known locations (14)(15). This database can then be queried with a given RSSI fingerprint (as measured by the user's smartphone), to obtain the estimated location of the user. In our experiments, we used the inverse mapping technique proposed in (23), and summarized as follows. Given a RSSI fingerprint, we find the k closest (under Euclidean metric) fingerprints in the database. The estimated location is then set to be the weighted average of the locations associated to these k signatures, where the weights are proportional to the sum of the RSSI of each signature (the weights sum up to 1). In our experiments, we used $k = 4$.

Calibration Calibration of the 3-beacon system was achieved using a simple and effective procedure. We created a polar grid centered around a specific location within the waiting/boarding area, with angular separation of 30 degrees, and radial separation of 0.5 meters (see Figure 4). Ten RSSI fingerprint measurements were taken at each location on the polar grid, with half of these measurement taken with the user holding the smartphone facing the center of the grid, and half while facing the opposite direction. The polar coordinates of each measurement

1 point were then converted to Euclidean coordinates defined on a local reference system, and
 2 finally to lat/long coordinates based on the known location of a nearby landmark, and the known
 3 orientation of a nearby street, used as reference.

4 For the bus stop equipped with a single beacon, we simply determined an RSSI
 5 threshold corresponding to a distance of approximately 5 meters from the beacon. Note that in
 6 this case, there is no need for accurate localization during the calibration phase; only the distance
 7 to the beacon (e.g. measured with a measuring tape) needs to be computed.
 8



9
 10 **FIGURE 4** An illustration of the polar grid of locations used to calibrate the beacon system.
 11

12 **Accuracy Assessment** The localization accuracy of the 3-beacon system was measured by
 13 walking with constant speed along a straight segment with known endpoints, and by recording
 14 start and stop time (which allows for accurate estimation of the user's location at any time during
 15 the walk). RSSI fingerprints were collected at a rate of one measurement per second, and
 16 location was estimated using the algorithm described above. Averaged over five data sets thus
 17 collected, we found a mean localization error of 3.4 meters. For comparison, the mean
 18 localization error of GPS (as measured in the same trials) was of 6.3 meters. It should be noted
 19 that while the localization accuracy from beacon RSSI fingerprint was consistent across trials,
 20 GPS resulted in errors varying from 3.2 meters to 13.3 meters (over different trials). We noted in
 21 our experiments that the localization accuracy provided by the beacon system is highly
 22 dependent on the user's orientation; this is likely due to the fact that if the user's body occludes
 23 view of the beacons, the measured RSSI drops, resulting in high error variance. It is also
 24 important to note that the accuracy of both GPS and beacons system depends on various
 25 conditions, such as the presence of nearby tall structures that can occlude view of the GPS
 26 satellites, or the presence of people or other occluders that can attenuate the radio signal from the
 27 beacons. For the case of the bus stop considered in this experiment, a canopy of redwood trees
 28 likely contributed to attenuation of GPS signal, while the presence of a large tree was responsible
 29 for attenuation of the signal in some spots within the bus stop area from the beacon marked as
 30 'C' as shown in Figure 3.
 31

32 *GPS vs. Beacon-Based Localization*

33 As discussed above, while universally available (in the outdoors), GPS provides, in general, a
 34 much lower localization accuracy than the BLE beacon system. For example, Figure 5 (a) shows
 35 a situation in which the user, while walking towards a bus stop (marked as 'A'), was out of range
 36 of the beacons in the bus stop, and thus could only rely on GPS. In Figure 5 (b), two localization
 37 estimates are shown, one from GPS and one from the beacons (the latter with a much smaller
 38 confidence radius). The GPS confidence radius was accessed via the Android Location class; it
 39 represents the radius of the smallest circle centered at the estimated location that is assumed to

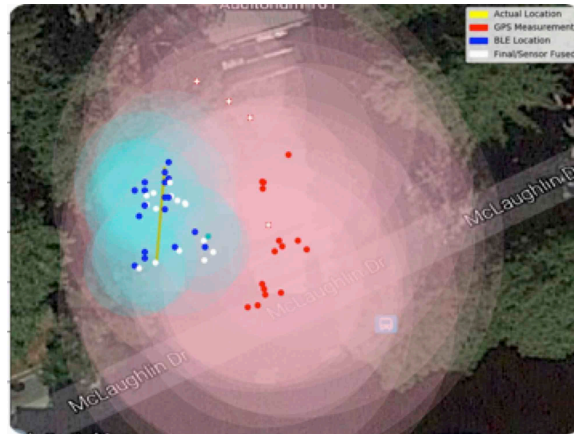
1 contain the true location with probability of 0.68. In order to obtain a similar measure of
 2 uncertainty for beacon-based localization, we regressed the variance of measured localization
 3 error on the sum of RSSI values in the fingerprint.
 4



FIGURE 5 Views generated by the web application in two different moments while the traveler was approaching the bus stop marked as ‘A’ in the map. The circles represent the confidence radius for localization provided by GPS or beacons.

The difference between the two systems’ accuracy is also clearly visible in Figure 6, showing a sequence of locations estimates for a user walking along the straight path depicted by a yellow segment. This figure shows by means of red dots the user’s location as measured by GPS, with associated circles of confidence. Blue dots represent location estimates (and associated confidence circles) based on the measured RSSI fingerprints. Note how these estimates better represent the actual path walked by the user (even though relatively large errors are occasionally observed).

The white dots in the figure represent location estimates obtained by statistical fusion of the two types of measurements. Specifically, these values are computed as the weighted average of GPS and beacon-based measurements, with weights inversely proportional to the variance of estimation. Fusion of GPS and beacon-based localization could be particularly useful in situations with low density of beacons (and thus poor associated accuracy). For example, in the case of a single beacon placed at a bus stop (which, as mentioned earlier, can be used only for proximity sensing), fusing information with GPS could allow users to estimate on which side of the beacon they find themselves on the sidewalk. This could be useful for a person (e.g. a blind traveler) to understand where to move in order to get closer to the beacon.



1
2
3 **FIGURE 6 Location estimates and confidence circles computed at different times while the**
4 **traveler was walking along the straight path represented by the yellow segment. Blue:**
5 **beacons; red: GPS; white: statistical fusion of localization from beacons and GPS.**
6
7

8 **Vehicle Arrival Notification**

9 Besides aiding in self-localization, BLE beacons can be placed inside a bus vehicle or train car,
10 allowing passengers waiting at a bus stop or platform to be notified upon arrival of a specific
11 vehicle. Thanks to the short discovery time (on the order of one second), notifications can be
12 produced while the vehicle is still approaching the stop, or as soon as it pulls over. Receiving
13 reliable information from the system that the vehicle that has just arrived is the correct one may
14 be very important for travelers with low or no vision, as well as for travelers who are uncertain
15 about the bus or train line to take. Another advantage of in-vehicle beacon placement is that the
16 user's app can sense the beacon throughout the ride, enabling the cloud system to ascertain that
17 he or she is riding the correct bus, and produce contextual information (e.g., notifications about
18 the upcoming destination stop). Note that, while it is possible in principle to identify if a traveler
19 is riding a certain bus line based on the GPS trace of his or her smartphone (24), this can be
20 difficult in the case of different lines sharing a common portion of the route. In contrast,
21 detection of a in-vehicle beacon over an extended period of time provides positive confirmation
22 that the user is riding that particular bus or train.

23 Two problems need to be addressed for successful vehicle arrival notification. The first
24 problem is disambiguation of situations when two (or possibly more) bus vehicles arrive at the
25 same time, with the user's smartphone in the transmission range of the beacons from both
26 vehicles. It is conceivable that the ID of the closest beacon could be determined on the basis of
27 the received RSSI; we plan to conduct future studies to confirm this hypothesis. The second
28 problem is the association between a beacon ID and a specific bus line. Beacons identify bus
29 vehicles; however, the same vehicle could be (and typically is) used for different lines in
30 different days. This calls for a mechanism to expose the vehicle's identity. For example, if the
31 transit agency provides a GTFS Real Time feed, this information could be embedded in the trip's
32 optional *VehicleDescriptor* field.
33
34
35



1
2
3 **FIGURE 7 A UC Santa Cruz campus bus vehicle that was equipped with a Kontakt.io BLE**
4 **beacon.**

5
6 **CONCLUSIONS**

7 We introduced RouteMe2, a cloud-integrated service designed to facilitate use of public transit
8 by those who may have difficulties with tasks such as identifying a bus stop or train platform,
9 finding the waiting/boarding area in the transit stop, ascertaining whether an incoming bus or
10 train is the desired one, and determining when to exit from the bus or train. Rather than
11 addressing each of these functionalities independently, RouteMe2 takes a global approach,
12 effectively operating as an end-to-end travel assistant. Importantly, it allows authorized
13 individuals (family members, caregivers) to monitor the traveler's progress in a registered trip.
14 RouteMe2 is supported by an infrastructure of inexpensive BLE beacons, placed at bus
15 stops/train platforms and inside bus vehicles and train cars.

16 A pilot implementation of the RouteMe2 system was deployed on the UC Santa Cruz
17 campus, demonstrating some of its most critical functionalities. Future work will include adding
18 access to available GTFS Real Time feeds for just-in-time re-routing (e.g., if a bus line is
19 delayed), improved accessibility of the mobile application (e.g., using VoiceOver for blind
20 traveler), and tests with a larger-scale deployment with partnering transit agencies.

21
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28

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