

1 **A Network-Sensitive Transport Modeling Framework for Evaluating Impacts of Network Disruptions**
2 **on Traveler Choices under Varying Levels of User Information Provision**

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1 **ABSTRACT**

2 There is considerable interest in the application of active traffic demand management (ATDM) and
3 traveler information systems strategies to mitigate the adverse impacts of congestion and network
4 disruptions. Such strategies and user information provision systems not only impact network
5 performance through the modification of traveler route choices, but also through changes in the entire
6 range of activity-travel choices such as activity generation, destination choice, mode choice, and time of
7 day choice. The simulation of the impacts of alternative strategies on network performance therefore
8 calls for the development and application of integrated modeling frameworks capable of reflecting the
9 entire slate of activity-travel pattern adjustments that may occur in response to changes in network
10 conditions and user information provision. This paper describes an integrated modeling framework
11 wherein an activity-based travel demand model and a dynamic traffic assignment model are tightly
12 coupled together with continuous information exchange between the models along the continuous time
13 axis. The framework is enhanced to accommodate the possible impacts of alternative user information
14 provision strategies on traveler choices and applied to a subregion in the Greater Phoenix metropolitan
15 area to demonstrate the sensitivity of the model to network disruptions under alternative information
16 provision scenarios. Model results are consistent with expectations and show that impacts of network
17 disruption are substantially mitigated in the presence of traveler information systems. Further, the
18 model results show that workers – who have more rigid work schedules and locations – are more greatly
19 impacted by network disruptions than non-workers who do not have such constraints. Integrated
20 modeling tools such as that described in this paper offer promise for evaluating emerging operational
21 and policy strategies aimed at influencing traveler choices.

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24 **Keywords:** Network disruption, traveler information systems, traveler choice simulation, travel demand
25 and network supply interaction, integrated travel models
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1 **INTRODUCTION**

2 Network disruptions refer to a class of events that alter the regular flow of traffic on one or more
3 roadway facilities. Network disruptions lead to a drop in capacity on the roadway element where the
4 event occurs, and cause delays, build up queues, and result in spillbacks on to surrounding links in the
5 network. Network disruptions may include planned events such as full roadway or lane closures to
6 accommodate work zones along a freeway segment or bridge section, or unplanned events such as
7 traffic crashes or roadway/bridge failures. The ability to model the impacts of network disruptions on
8 travel demand and traffic flow is important for a number of reasons. First, in the context of unplanned
9 network disruptions, understanding the impact of such events and associated delays allows for the
10 planning of emergency response services. Emergency response services can be routed and delivered so
11 that crisis teams tend to incidents as quickly as possible and alleviate the impact of disruptions. Second,
12 modeling the impact of network disruptions allows for estimating the changes in activity-travel demand
13 along both the space and time dimensions that may result due to such events. Such an understanding
14 would allow professionals to devise traveler information systems and routing strategies that minimize
15 adverse impacts on people’s activity-travel schedules.

16 The effects of network disruptions may be simulated using a variety of transport modeling tools.
17 However, the following key considerations play a key role in the development of appropriate modeling
18 frameworks capable of simulating the dynamics associated with network disruptions:

- 19 • The model system should be able to account for the effects of information provision while
20 accurately representing the spatial and temporal scales of the information that the different sources
21 seek to provide. For example, there are a number of outlets that provide information about real-
22 time network conditions such as Google maps that provide information about traffic conditions only
23 along major freeway and arterial corridors, radio traffic reports that provide more spatially
24 aggregate information about incidents, variable message signs on selected freeway corridors, and
25 commercial services that provide real-traffic information for a fee. These are but a few of the
26 Advanced Traveler Information Systems (ATIS) technologies capable of providing system-wide
27 information based on prevailing network conditions. However, each of these technologies differs in
28 the spatial and temporal extents of coverage, and these spatio-temporal dimensions need to be
29 effectively captured in a model used to simulate traveler response to user information systems
30 under network disruptions.
- 31 • There are different decision paradigms that characterize how individuals’ process information during
32 a network disruption. When provided with information prior to embarking on a trip (pre-trip),
33 individuals may alter their destination or mode, or completely forego the activity based on network
34 conditions. On the other hand, individuals that are en-route to an activity destination may change
35 their route, destination, or skip the activity altogether. The system modeling tool should be able to
36 capture both pre-trip and en-route decision processes with respect to how the activity-travel events
37 are planned and executed.
- 38 • Finally, trips should not be considered in isolation when evaluating impacts of network disruptions.
39 Trips result from an individual’s desire to engage in activities, subject to his/her knowledge about
40 the network. Prevailing network conditions and the choices made in response to information
41 provision (or lack thereof) affect subsequent activity-travel engagement decisions in the latter part
42 of the day. For example, if an individual is oblivious to an incident along a planned route, then he or
43 she may experience congestion and arrive late at a destination. This delay may have cascading
44 effects resulting in a modification of subsequent activity-travel engagement patterns. Upon arriving
45 late at the activity destination, the individual may adjust the duration of the activity, or keep the
46 duration of the activity constant, but alter subsequent activity-travel episodes. The same individual,
47 when provided with information about network conditions, may skip the planned activity or choose
48 a different destination to pursue the same activity; both of these decisions could impact subsequent

1 activity-travel engagement decisions. It is necessary for models to provide a holistic accounting of
2 individuals, their behaviors and interactions, and the daily activities and trips they wish to pursue.
3 This last consideration also has implications for drawing inferences regarding quality of life impacts.
4 If trips are considered in isolation, then computed quality of life indices may not be representative
5 of the entire daily activity-travel patterns of individuals, and could lead to inaccurate policy
6 inferences.

7 In this paper, a framework for modeling network disruptions which allows for an accurate
8 representation of activity-travel engagement, network dynamics, and the interplay between these two
9 components is presented. The framework, called SimTRAVEL (Simulator of Transport, Routes, Activities,
10 Vehicles, Emissions, and Land) combines a land use - travel demand model system generating activity-
11 travel engagement decisions with a traffic simulation model which routes and simulates the movement
12 of vehicles on the network. A prototype of SimTRAVEL has been developed by integrating
13 microsimulation-based models of land use (UrbanSim; Waddell et al. 2008), travel demand (OpenAMOS
14 – an open-source activity-based travel demand model system), and network dynamics (MALTA - Multi-
15 Resolution Assignment and Loading of Traffic Activities; Chiu and Villalobos, 2008). The prototype is
16 designed with a view to accurately capture the interactions and constraints that people experience as
17 they pursue their activity-travel agendas. The prototype, described previously in Pendyala, et al (2011),
18 is used to model the impacts of an unplanned network disruption on a major freeway corridor. A
19 comprehensive analysis is conducted to assess the impact of the network disruption on activity-travel
20 engagement patterns under a variety of traveler information provision scenarios.

21 The remainder of this paper is organized as follows. In the next section, a discussion on
22 modeling network disruptions under user information provision is furnished. Then, the integrated
23 modeling framework and the SimTRAVEL prototype that has been developed are described. In the
24 fourth section, the case study and the different scenarios that are evaluated are described. In the fifth
25 section, results from the application of the prototype to model different network disruption scenarios
26 are presented. Finally, the paper offers a discussion of the implications of the results along with some
27 concluding thoughts.

28 **MODELING NETWORK DISRUPTIONS UNDER USER INFORMATION**

29 There is a rich body of literature in the transportation domain focusing on better understanding the
30 impact of network disruptions. The literature may be broadly classified into two themes, namely,
31 measuring activity-travel behavior changes in response to network perturbation and user information
32 provision, and the development of system modeling tools for simulating network disruptions and
33 evaluating different policies and solutions to alleviate the impacts of network disruptions. Along the
34 first theme of understanding the activity-travel behavior dynamics associated with network disruptions,
35 Zhu et al. (2010) use measurements from the field to quantify the impacts of the collapse of the I-35W
36 bridge, which spans the Mississippi River in Minneapolis, on traffic flows in the region and travel
37 behavior patterns. Yun et al. (2011) explore travel behavior changes exhibited by individuals in response
38 to a planned network disruption – a nine-week reconstruction project of Interstate 5 (I-5) in
39 Sacramento, California. They conduct surveys of commuters to measure changes in traveler behavior
40 brought about by the planned highway closure. The study presents results from the first two surveys
41 and highlights the types of behavioral changes exhibited by individuals. Other studies include those by
42 Chang and Nojima (2001), Zhu et al. (2010), and Kamga et al. (2011) who examine impacts of unplanned
43 network disruptions, and that by Clegg (2007) which focuses on the impacts of planned network
44 disruptions. While these studies are dedicated to understanding the impacts of network disruptions on
45 traveler behavior, they are rather limited in scope – examining specific dimensions of activity-travel
46 engagement behavior, or focusing exclusively on specific demographic segments. There are very limited
47 studies that examine impacts on the entire range of activity-travel choices in response to network
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1 disruptions, as well as network dynamics arising from the deployment of one or more disruption
2 mitigation strategies. Several studies highlight the data collection issues and other challenges faced by
3 researchers in understanding the behavioral changes exhibited by individuals in response to network
4 disruptions.

5 Within this domain, there have been several studies aimed at understanding how traveler
6 information provision may impact travel behavior in the event of a network disruption. Levinson (2003)
7 explores the benefits gained by individuals due to in-vehicle navigation systems that provide real-time
8 traveler information about both recurring and non-recurring congestion. Kraan et al. (2000) present
9 results from a stated preference survey conducted to understand traveler responses to advanced
10 traveler information systems. The study was limited to shopping trips and survey respondents were
11 afforded the flexibility to altering their destinations and desired arrival time. However, the study did not
12 consider a number of other possible activity-travel adjustments that individuals may make, such as
13 skipping activities, re-scheduling activities, and altering route. Liu and Mahmassani (1998) present
14 results from an advanced dynamic interactive traveler simulator experiment exploring the different
15 types of behaviors exhibited by individuals in response to advanced traveler information systems. These
16 studies identify the need for additional research on the types of behaviors exhibited by individuals when
17 presented with user information, and the role of socio-economic and demographic variables on the
18 behaviors exhibited.

19 In the modeling tool development domain, previous research has largely focused on modeling
20 dynamics in network operations under a variety of mitigation strategies while assuming that the travel
21 demand itself is exogenous. He and Liu (2011) present an approach for modeling network day-to-day
22 network conditions after the onset of an unexpected network disruption. They present calibration and
23 validation results using data collected after the I-35W bridge collapse. As mentioned earlier, these
24 research efforts do not model the entire range of activity-travel decisions that may be altered in
25 response to network disruptions. Sundaram (2002) presents a framework for modeling network
26 disruptions with a view to capturing their impacts on activity-travel behavior. However, the model
27 system constitutes a hybrid model with a traditional four-step model travel demand as opposed to a
28 full-scale microsimulation model of activity-travel behavior that provides for a more accurate
29 representation of underlying behaviors, interactions, and constraints characterizing travel. Jenelius et al.
30 (2011) present a model of departure time choice in response to network conditions in the context of trip
31 chains formed by individuals. The study is however limited to trip chains with three activities i.e. two
32 trips, and associated departure time choices. The model presented in their study does not
33 accommodate changes in other activity-travel choices such as activity type choice (activity generation),
34 destination choice, and number of stops.

35 With respect to modeling the impacts of traveler information systems technologies on network
36 performance under various network perturbation scenarios, Al-Deek et al. (1998) consider three
37 different types of travelers. The three types of travelers are those that do not have ATIS, those that
38 have delayed ATIS in the form of radio or other technologies, and those that have real-time ATIS. They
39 aimed to study the impact of different technology configurations and analyze the traveler and system
40 benefits derived from varying levels of technology penetration. Paz and Peeta (2008) present a
41 framework for generating traffic routing strategies by accounting for driver's likely behavior in response
42 to path recommendations provided by a user information system. While the model systems presented
43 in prior research provide platforms to evaluate the impact of technology on network path choices and
44 performance, the models do not consider activity-travel engagement decisions (scheduling and re-
45 scheduling and other dynamics) that individuals may alter from the use (or lack thereof) of traveler
46 information technologies.

1 **A NETWORK-SENSITIVE ACTIVITY-TRAVEL SIMULATION FRAMEWORK**

2 A key consideration in modeling the impacts of network disruptions is that the travel modeling and
3 simulation framework should be able to capture the activity-travel scheduling and rescheduling behavior
4 exhibited by individuals in response to network delays. This calls for the integration of an activity-based
5 travel demand model with a dynamic traffic assignment model within a unified framework that
6 maintains consistency in activity-travel choice processes as patterns evolve over the course of a day.
7 Considering the dynamics associated with network disruptions, the framework should allow the activity-
8 based travel demand and traffic assignment model systems to communicate with each other along the
9 continuous time axis. Only a framework where the demand and supply model systems constantly
10 exchange information is capable of truly capturing activity-travel demand and network supply dynamics
11 in the presence of network disruptions and various configurations of traveler information provision.

12 An approach that has been proposed to integrate activity-travel demand and network supply
13 models is to run the models sequentially (Lin et al. 2008; Kitamura et al. 2005). In such a configuration,
14 each of the model systems is run separately and linked together sequentially through input-output data
15 flows and feedback loops. However, the sequential approach cannot be used to model the impacts of
16 network perturbations because the framework does not provide for constant communication between
17 the model systems along the continuous time axis. The modeling of network disruptions calls for an
18 event-based approach to integrating the activity-based travel demand and dynamic traffic assignment
19 model systems.

20 An event-based model integration framework with tight coupling across the model systems was
21 presented by Kitamura et al. (2008) and more recently operationalized by Pendyala et al. (2011). Figure
22 1 presents the integrated model framework that has been operationalized within SimTRAVEL. In each
23 minute of the day, the demand model simulates activity-travel engagement decisions of all individuals.
24 Information from the demand model, including, origin, destination, mode, and vehicle attributes for all
25 trips that need to be executed on the network, is then passed to the dynamic traffic assignment model
26 for identifying the route and simulating the movement of vehicles on the network. Once the trips arrive
27 at their destination, the traffic assignment model passes back the arrival (time) information to the
28 demand model to simulate activity-travel engagement decisions in subsequent time steps. The activity-
29 travel demand model simulates activity schedules in subsequent time steps based on actual arrival times
30 experienced by travelers. Thus, if a traveler is delayed in arriving at a destination due to congestion on
31 the network, then subsequent activity engagement decisions will be affected. These types of impacts
32 are effectively captured in the event-based paradigm implemented within SimTRAVEL. This process
33 proceeds for all 1440 minutes within a day and activity-travel schedules for an entire day are generated.
34 Network conditions obtained at the end of the daily simulation are then fed back to re-simulate activity-
35 travel schedules and the process is repeated until convergence in the network conditions is achieved,
36 i.e., the network inputs that are used to simulate activity-travel patterns and the network conditions
37 resulting from the activity-travel patterns are the same. It should be noted that, within the basic design
38 of the framework, network skims from a previous iteration are being used by the demand model to
39 simulate activity-travel choices (destination choice, mode choice, activity type choice) but actual
40 network conditions prevailing within the current iteration at the appropriate time step are used to
41 determine traveler arrival times at destinations.

42 The event-based framework presented in Figure 1 lends itself to modeling network disruptions
43 and can be used for understanding impacts of network disruptions on activity-travel engagement
44 decisions. In the context of modeling network disruptions, however, there are two key features that an
45 integration framework needs to support. First, the actual arrival times need to be fed back to the travel
46 demand model to simulate activity-travel engagement decisions in the subsequent time interval. This is
47 already accommodated in the basic framework described earlier. Second, the model system should be
48 able to use travel skims (travel times) of the current iteration (as opposed to the previous iteration) in

1 simulating activity-travel choices. The basic framework described earlier needs to be modified to
2 accommodate this capability so that the modeling of information provision impacts can be
3 accomplished. In order to evaluate the impacts of real-time traveler information systems, the model
4 framework should be able to pass network conditions prevailing after the onset of an incident to the
5 travel demand model so that the simulated activity-travel engagement patterns are a reflection of the
6 network conditions that prevail at the time (as opposed to “expected” travel times derived from a
7 previous iteration). The prevailing network conditions should also be used in identifying time-dependent
8 shortest paths for travelers who may be seeking to avoid the congested (disrupted) portion of the
9 network. Essentially, the framework presented in Figure 1 can accurately capture the first
10 consideration, i.e., adjusting activity-travel scheduling behavior in response to arrival information.
11 However, the framework cannot simulate information provision, i.e., the framework does not allow the
12 utilization of prevailing network travel times for simulating activity-travel choices and routing decisions
13 in the subsequent time period(s) of the day after the onset of the disruption. Therefore, the event-based
14 framework presented in Figure 1 is further enhanced for this study so that it can be used to evaluate the
15 impacts of various information provision scenarios, which is very important application area for such a
16 dynamic integrated modeling tool.

17 Figure 2 presents a revised event-based framework for integrating the activity-based travel
18 demand and dynamic traffic assignment models that supports modeling information provision under
19 network disruption. The model system proceeds in the same minute-by-minute fashion presented in
20 Figure 1, where converged base year link travel times (L_{base}) from the previous iteration are used from
21 the start of day until the onset of the disruption ($t = a$) and again from the time that the disruption is
22 cleared ($t = b$) until the end of day. However, for the time period between onset and clearing of the
23 disruption ($a \leq t \leq b$), the linkage between the travel demand model and the traffic assignment and
24 simulation model is modified as follows:

- 25 1. At the end of every simulation interval (t), the dynamic traffic assignment model replaces the
26 expected link travel times (L_{base}) from the previous iteration with the existing travel times (L_t) on the
27 network for the current and all subsequent intervals because that is the best estimate of prevailing
28 and future network conditions after the onset of an incident.
- 29 2. The new link travel times (L_t) by time of day are used to generate origin-destination travel time
30 matrices (OD_t), i.e., time varying skim files, for use in the travel demand model.
- 31 3. The dynamic traffic assignment model passes the travel time matrix (OD_t), reflecting prevailing
32 conditions, along with all trips that have arrived at their destination, to the demand model so that
33 activity-travel engagement decisions for the subsequent time interval may be simulated.
- 34 4. The travel demand model, in turn, passes trips that need to be loaded on the network, which
35 travelers have chosen to undertake based on information about the prevailing network conditions
36 (OD_t). In response to the prevailing (delayed) conditions, people may choose alternate destinations,
37 or may just choose to proceed early to their next fixed/mandatory activity (e.g., work) because they
38 know it will take longer to get to the fixed activity.
- 39 5. Once the trips are received by the dynamic traffic assignment model, routes are identified using
40 prevailing conditions (L_t) as the expectation of the network for all subsequent time intervals. The
41 traffic assignment model then loads and routes/simulates the trips through the network.
- 42 6. The simulation time step is incremented ($t = t+1$) and the process (Steps 1 - 5) is repeated until the
43 incident is cleared.
- 44 7. Once the incident has cleared, the base year converged network conditions by time of day are used
45 once again to simulate activity-travel engagement and routing decisions.

46 The process described above is represented in the flowchart depicted in Figure 2. The flowchart
47 presented in Figure 2 offers a robust framework for modeling traveler choices under alternative
48 information provision scenarios in the event of a network disruption. As mentioned earlier, the time-

1 sensitive activity-travel simulation framework presented in this section has been operationalized as part
2 of a larger research effort by Pendyala et al. (2011). The prototype model system, dubbed SimTRAVEL
3 (Simulator of Transport, Routes, Activities, Vehicles, Emissions, and Land), integrates microsimulation-
4 based model systems of land use, activity-based travel demand, and dynamic traffic assignment
5 consistent with the framework presented in Figure 1. The land use model system employed in the
6 prototype is UrbanSim, the travel demand model system employed is OpenAMOS (Open-source Activity
7 Mobility Simulator) and the dynamic traffic assignment model implementation is MALTA (Multi-
8 Resolution Assignment and Loading of Traffic Activities) (Pendyala et al. 2011). For purposes of this
9 study, the SimTRAVEL prototype was enhanced to incorporate additional feedback between the travel
10 demand and dynamic traffic assignment model systems as necessitated by the framework presented in
11 Figure 2 so that network perturbations can be modeled accurately under information provision
12 scenarios.

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14 **CASE STUDY APPLICATION**

15 The primary objective of this effort was to demonstrate the applicability of the enhanced SimTRAVEL
16 prototype to evaluate a network disruption scenario under alternative information provision strategies.
17 The applicability and the different features of the prototype are demonstrated by modeling three
18 disruption scenarios with varying levels of information provision and subsequently analyzing the activity-
19 travel dynamics under these scenarios. The three scenarios evaluated in this study are described below:

- 20 • *No disruption*: In this scenario, no incident occurs and base line conditions prevail. The scenario
21 serves as a baseline against which the other two scenarios are compared and analyzed.
- 22 • *No information provision*: In this scenario, a planned network disruption is introduced; however
23 people are assumed to be oblivious to the disruption and go about planning and executing their
24 activity-travel patterns based on their previous experience (expectations) of network conditions.
25 Individuals are assumed to be using an expectation of the network conditions in making decisions
26 about activity-travel engagement and route choices. The assumption of no information provision
27 and complete ignorance of the network conditions may be unreasonable given the ubiquity of
28 different types of information media and communication technologies (e.g., 511 systems, radio, and
29 real-time traffic data like Google Maps) which afford individuals the ability to learn about real time
30 traffic conditions. In this scenario, individuals are assumed to have no information during the trip
31 planning stage, i.e., prior to actually embarking on a trip. Once the individuals are on the network,
32 they experience delays due to the disruption, actual arrival times reflect the impact of the
33 disruption, and travelers make adjustments to their activity-travel engagement choices in
34 subsequent time periods in response to delays experienced due to the disruption. However, every
35 activity-travel (demand) engagement decision (pre-trip) itself is made based on expectations of
36 travel times (i.e., converged travel times from the base model run with no disruption). Changes and
37 adjustments in activity-travel patterns emerge in subsequent time periods because people arrive at
38 destinations considerably delayed, and face subsequent time constraints that result in activity travel
39 patterns different from those that would have emerged had there been no disruption on the
40 network.
- 41 • *Full information provision*: In this scenario, travelers that are already on the network follow their
42 planned routes even after the onset of the network disruption (i.e., there is no enroute switching).
43 However, individuals that are about to embark on a trip (pre-trip stage) are assumed to be aware of
44 the incident and the prevailing network conditions. The activity-travel engagement and routing
45 decisions of these individuals are based on the prevailing network conditions and not based on
46 expected conditions of the network that they generally experience. Thus, while travelers already on
47 the network at the onset of the disruption will inevitably be considerably delayed and need to adjust
48 their subsequent activity-travel schedules in response to delayed arrival times, those that embark on

1 trips after the onset of the disruption will be able to plan their trips with full knowledge of network
2 conditions. They can presumably choose alternative destinations (for non-mandatory activities),
3 activity types, and routes to best plan around and avoid adverse impacts of the network disruption.
4 Every traveler, including those that were already on the network at the onset of the disruption, is
5 assumed to have full knowledge of prevailing network conditions at the pre-trip planning stage.
6 Thus, a traveler who was on the network at the onset of the disruption and arrived late at his or her
7 destination is now fully aware of the network conditions when planning the next trip.

8 The no information provision and full information provision scenarios comprise network
9 disruptions under two rather extreme levels of information provision, with some simplifications on the
10 representation of activity-traveler choices. In reality, traveler awareness of network conditions is likely
11 to fall between these two extremes modeled in this study. Moreover, in the real-world, travelers are
12 likely to change paths or destinations en-route; however, the current modeling framework in
13 SimTRAVEL does not accommodate en-route adjustments to routes and destinations. People who are
14 already on the network at the onset of the disruption must proceed to the planned destination on the
15 planned route (based on expected travel times). Nevertheless, modeling and analyzing the two
16 information provision scenarios described earlier demonstrates the abilities of a dynamic integrated
17 model system in simulating the impacts of network disruptions under alternative information provision
18 strategies while also providing some indication of the range of impacts on activity-travel engagement
19 behavior that may result from a network disruption. The study sheds light on the impact of information
20 provision during network disruption on traditional measures of network conditions (total trips, and
21 delays) and also on activity-travel engagement behavior (trip lengths, trip durations, trip rates, and daily
22 time allocation).

23 The case study area consists of three cities (Chandler, Gilbert, and Queen Creek) in the
24 southeast region of Maricopa County (Greater Phoenix) in Arizona (Figure 3). There are about half a
25 million people residing in about 150,000 households in the three city area. The demand for the three
26 city region is generated using the full-scale microsimulation model in SimTRAVEL while the demand for
27 the rest of the Maricopa region was captured by creating trip lists from origin-destination tables
28 produced by the existing four-step travel model. This enabled modeling the impact of network
29 conditions while accounting for congestion on the roadway network in the model region. Essentially,
30 about two million trips were simulated by the activity model within SimTRAVEL for the residents of the
31 three-city region, while the remaining 13 million trips in the region were derived from the origin-
32 destination tables of the calibrated Maricopa Association of Governments (MAG) four-step travel
33 demand model. A planned network disruption was introduced by dropping the lane capacity of a
34 section of the freeway that runs through the middle of the three city region and connects it to the rest
35 of the Maricopa (Greater Phoenix) region. The simulation considers an incident situation wherein only
36 one lane in each direction of the freeway segment is operational (as opposed to the usual three general
37 purpose lanes in each direction) and other lanes are closed to clear the incident. The incident was
38 assumed to start at 7:00 AM and end at 10:00 AM. Within the simulation, it is assumed that the onset
39 of the disruption and the clearing of the disruption are both instantaneous; in other words, the
40 simulation does not reflect the more gradual onset of delays and congestion, and dissipation of queues,
41 that are observed in reality. This simplification was done for operational convenience and future
42 development efforts will focus on more realistically capturing such phenomena.

43 The time period was chosen to reflect peak travel demand generated by individuals residing in
44 the three city region. In Figure 3, the stretch of freeway that is disrupted is on the Santan Loop 202 that
45 is a vital freeway corridor for the southeast region of Maricopa County. Residents of the three cities of
46 Chandler, Gilbert, and Queen Creek rely heavily on this freeway corridor to commute and access other
47 shopping and social-recreation destinations. Due to data limitations, the land use simulation model
48 (UrbanSim) had to limit work and school locations of residents in the three city area to fall within the

1 boundaries of the three city area. In reality, many residents of the three city area have work locations
2 outside the three-city area (say, in Tempe, Mesa, Phoenix, Scottsdale); by limiting work locations to fall
3 within the localized region, it is likely that the simulation is not fully replicating travel patterns
4 undertaken by workers in the region on a weekday. However, no such limitation exists for non-work
5 trips; non-work trip destinations may fall outside the boundaries of the three-city area. As the limitation
6 only impacts a subsample of workers (whose work locations are in reality beyond the boundaries of the
7 three-city area) in the context of their work trips and associated trip chains, and with 13 million trips
8 derived from the 2009 validated four-step travel model for the region, it is not surprising that the
9 SimTRAVEL model system was found to validate very well against observed activity-travel patterns in the
10 2009 National Household Travel Survey (Pendyala et al. 2011).

11 **MODEL APPLICATION RESULTS**

13 Before running the three scenarios, the SimTRAVEL prototype was run iteratively to obtain a set of
14 stable and converged base year time-dependent origin-destination skim matrices. This set of network
15 travel times from the base year simulation run was then used to launch the three scenarios. A variety of
16 demand characteristics were examined to assess the impacts of network disruptions under alternative
17 user information provision strategies. The number of trips generated in the baseline no disruption case
18 is 14,320,888 trips. In the full information scenario, 14,321,746 trips were generated, whereas in the no
19 information scenario a total of 14,317,790 trips are produced. The number of trips generated in the no
20 information scenario is the smallest and this is consistent with expectations. In the no information
21 scenario, people are presumably planning trips and selecting routes without any knowledge about the
22 incident. As a result, they experience higher delays and spend more time on the network, which will in
23 turn affect their time constraints for subsequent activity-travel engagement decisions. Available time to
24 pursue activities will shrink leaving individuals with less time to engage in any non-fixed activities
25 (relative to the baseline no disruption case). It is interesting to note that the number of trips generated
26 in the full information scenario is higher than the no disruption baseline scenario by about 858 trips. It
27 is possible that there are shadow effects at play here. While travelers who are already en-route are
28 stuck in congestion, links downstream of the bottleneck and neighboring roadways may be experiencing
29 free flow conditions relative to the baseline no disruption case due to shadow effects. As a result,
30 individuals (with full information) planning their trips will alter destinations and routes and take
31 advantage of the higher level of service on other links (in the shadow of the disruption). These
32 individuals arrive earlier than expected, and have more time to subsequently engage in activities and
33 travel. This phenomenon may be contributing to the slight increase in activity and travel engagement in
34 the full information provision case. The total vehicle miles traveled (VMT) in the full information
35 scenario is also about 10125 miles higher than the base scenario. The higher VMT may result from two
36 phenomena. First, as mentioned earlier, there is a slight increase in the number of trips pursued in the
37 full information case. Second, individuals are likely selecting alternative routes (e.g., using surface
38 streets) to avoid the section of the freeway affected by the incident to get to their activity locations,
39 because they have complete knowledge about the incident and the network conditions; these
40 alternative routes may be longer in distance (but shorter in time because of the network disruption on
41 the freeway section) resulting in the higher VMT.

42 Within the scope of this paper, it is not possible to provide an exhaustive set of results.
43 Therefore, two illustrative sets of results are presented in this paper. Figure 3 shows the shift in trip
44 duration (length) distributions for workers under the two information provision scenarios (relative to
45 the baseline case). The differences between the two scenarios are quite readily apparent. In the no
46 information provision case, the number of trips of shorter length drop dramatically with the level of
47 reduction decreasing with increasing trip duration. For example, the largest drop is seen in trips of 10-20
48 minutes duration and the smallest drop is seen in trips of 50-70 minutes duration. Conversely, there is

1 an increase in the number of trips of long duration. In the absence of information, workers are stuck in
2 congestion, considerably delayed, and the trip length distribution shifts considerably towards longer trip
3 lengths (durations). When there is full information provision, on the other hand, the impacts are far less
4 dramatic. There is a decrease, albeit smaller, in the number of very short trips, which suggests that
5 travelers cannot fully escape the impacts of the network disruption even when provided full
6 information. Some travelers who are already en-route when the disruption occurs will be delayed
7 (experiencing longer trip lengths) and those who begin trips during the disruption period may not be
8 able to fully avoid the critical freeway section in its entirety even though they are aware of the
9 disruption. There is an increase in the number of trips of medium duration, and a series of slight
10 increases in the long duration trip lengths (but far less dramatic than in the no information provision
11 case). These results are consistent with expectations and demonstrate the ability of the model system
12 to mimic shifts in travel patterns under alternative disruption and information provision scenarios.
13 Similar trends are seen for non-workers (figure not shown), except that the shifts are less pronounced
14 presumably because non-workers have greater flexibility and discretion in planning and executing trips
15 (with respect to timing, destination, and participation) in contrast to workers who have less degrees of
16 freedom due to rigid work schedules and locations.

17 Table 1 shows the daily time allocation to trips and activities across all individuals in the
18 simulation (i.e., the half-million people whose trips and activities were simulated within SimTRAVEL)
19 under the various scenarios (in units of person days). The trends are once again readily apparent. The
20 time spent on activities (i.e., at activity locations) drops quite dramatically in the no information
21 provision scenario relative to the full information provision scenario. For workers, the largest reductions
22 in activity time occur at home and at work, presumably because of the impact of the freeway disruption
23 on commute travel. There are reductions in maintenance and discretionary activity duration as well. In
24 the case of full information provision, there is a drop in work activity duration (presumably because
25 workers cannot fully escape the disruption in the context of commuting to a fixed work location), but
26 the drop is considerably less than in the case of no information provision. There is a large decrease in
27 maintenance activity duration, perhaps because workers decide to forego these activities, or shorten
28 these activities in duration, in the presence of full information about network conditions. As a result,
29 workers actually gain a little more time at home, and the overall reduction in out-of-home activity
30 duration is substantially smaller than in the no information provision case. Similar trends are seen for
31 non-workers except that the impacts on non-workers are substantially smaller. Non-workers
32 presumably have greater degrees of freedom and spend less time out of home to begin with – thus
33 dampening the impact of the disruption.

34 With respect to travel durations, similar impacts are observed. Workers experience the greatest
35 increase in travel time expenditure when there is no information provision. In particular, travel to and
36 from work increases considerably, presumably because of the need to use the freeway section for
37 commuting. In the presence of full information, there is some impact on travel time expenditure
38 (particularly for commute), but the impact is far less than in the case with no information provision. In
39 the case of non-workers, once again the impacts are dampened relative to workers with increases in
40 travel time considerably higher when there is no information provision compared to the scenario of full
41 information provision. In the case of full information provision, the travel time expenditure change is
42 quite modest – with slight decreases in travel time for maintenance and other trips. It is possible that
43 non-workers are choosing destinations and routes completely beyond the influence area of the
44 disruption for these activities, and save travel time expenditure (by choosing closer destinations and/or
45 benefiting from shadow effects). It is worth noting that, in all cases, consistency is maintained as the
46 aggregate reduction in activity duration is virtually identical to the aggregate increase in travel time
47 expenditure.

1 Overall, it can be seen that the model system is capable of reflecting the impacts of network
2 disruptions on activity-travel patterns and network dynamics under alternative information provision
3 strategies.

4 Given the small magnitude of changes noted in the metrics, one could argue that the changes
5 are just an artifact of the stochasticity associated with the microsimulation-based demand model. While
6 the stochasticity in the demand model does contribute to predicted differences across model runs,
7 there should not be any trends in the results if the differences are purely stochastic in nature. While
8 some of the changes observed may be attributed to stochasticity, the clear trends observed in the
9 results suggest that the differences are indeed caused by the altered input conditions (i.e., the level of
10 information provision).

11 **CONCLUSIONS**

13 This paper describes a research effort aimed at implementing a microsimulation-based integrated
14 modeling framework capable of simulating the impacts of network disruptions under alternative user
15 information provision strategies. The model system, called SimTRAVEL (Simulator of Transport, Routes,
16 Activities, Vehicles, Emissions, and Land), employs a paradigm wherein there is constant information
17 exchange between the activity-travel demand model and the dynamic traffic simulation model so that
18 activity-travel patterns evolve in response to actual network conditions experienced by travelers. The
19 scenario analysis undertaken in this study provides key insights into the impacts of network disruptions
20 on time use and travel behavior under different levels of information provision. The study demonstrates
21 the feasibility of applying an integrated dynamic model system to capture the complex behavioral
22 decisions and adjustments that travelers make in response to network conditions in the presence and
23 absence of information.

24 As part of this research effort, the original SimTRAVEL framework is modified to reflect
25 knowledge of prevailing network conditions on the part of travelers in the presence of user information
26 systems. The framework, when applied to three different scenarios, is found to offer plausible results
27 consistent with expectations. The model system is applied to a baseline case (no network disruption), a
28 no information provision scenario (with network disruption), and a full information provision scenario
29 (with network disruption). An examination of model outputs shows that impacts on activity-travel
30 patterns are more dramatic in the no information provision scenario (as expected) with a drop in
31 number of activities/trips and activity durations, and a corresponding increase in trip durations and daily
32 travel time expenditures. The differences are larger for workers, presumably because they have rigid
33 work schedules and locations that preclude them from being able to fully escape the effects of the
34 disruption even in the presence of full information. In all cases (for both workers and non-workers),
35 impacts are substantially less dramatic in the presence of full information.

36 This study offers promising results demonstrating the ability of the dynamic integrated model
37 system to simulate the impacts of operational active traffic demand management (ATDM) and traveler
38 information systems strategies under alternative network conditions. Future work should focus on
39 enhancing the effort along a number of lines. First, there is a need to more accurately quantify the
40 extent to which results are affected by stochasticity in the simulation model runs. By isolating the
41 stochasticity effects, it would be possible to more accurately predict the actual impacts of network
42 perturbations and information provision strategies on activity-travel patterns. Second, the model
43 system needs to be enhanced to avoid the instantaneous onset and clearing of the network disruption.
44 In the real world, there will be a more gradual build-up of congestion and queues at the onset of the
45 disruption and a gradual dissipation of queues upon reopening all lanes to traffic. By integrating
46 queuing models of traffic flow that simulate formation and dissipation of shockwaves at bottlenecks, the
47 model system can be made more reflective of real-world conditions experienced by travelers. Finally,
48 the model system needs to be applied on a full regional scale as opposed to a smaller subregion as was

1 done in this study. This would allow a more complete validation of the predictions offered by the model
2 system.

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8 9 **REFERENCES**

- 10 Al-Deek, H.M., Khattak, A.J., and Thananjeyan, P. (1998). A Combined Traveler Behavior and System
11 Performance Model with Advanced Traveler Information Systems. *Transportation Research A*,
12 32(7), pp. 479-493.
- 13 Chang, S. and Nojima, N. (2001). Measuring Post-Disaster Transportation System Performance: The 1995
14 Kobe Earthquake in Comparative Perspective. *Transportation Research A*, 35(6), 475-494.
- 15 Chiu, Y.-C. and J.A. Villalobos (2008). The Anisotropic Mesoscopic Simulation Model on the Interrupted
16 Highway Facilities. *Presented at the Symposium on the Fundamental Diagram: 75 Years*
17 *(Greenshields 75 Symposium)*, Woods Hole, MA.
- 18 Clegg, R. (2007). Empirical Studies on Road Traffic Response to Capacity Reduction. *Transportation and*
19 *Traffic Theory 2007: Papers Selected for Presentation at ISTTT17*, Elsevier Science, 155-178.
- 20 He, X. and Liu, H.X. (2011). Modeling the Day-to-Day Traffic Evolution Process After an Unexpected
21 Network Disruption. *Transportation Research B*, 46, 50-71.
- 22 Jenelius, E., Mattsson, L., and Levinson, D. (2011). Traveler Delay Costs and Value of Time with Trip
23 Chains, Flexible Activity Scheduling and Information. *Transportation Research B*, 45(5), 789-807.
- 24 Kanga, C.N., Mouskos, K., and Paaswell, R.E. (2011). A Methodology to Estimate Travel Time Using
25 Dynamic Traffic Assignment (DTA) Under Incident Conditions. *Transportation Research C*, 19(6),
26 1215-1224.
- 27 Kitamura, R., Kikuchi, A., Fujii, S., and Yamamoto, T. (2005). An Overview of PCATS/DEBNetS
28 Microsimulation System: Its Development, Extension, and Application to Demand Forecasting. In
29 R. Kitamura and M. Kuwahara (eds.) *Simulation Approaches in Transportation Analysis: Recent*
30 *Advances and Challenges*, Springer, New York, 371-399.
- 31 Kitamura, R., Kikuchi, A. and Pendyala, R.M. (2008). Integrated, Dynamic Activity-Network Simulator:
32 Current State and Future Directions of PCATS-DEBNetS. Presentation at the Second TRB
33 Conference on Innovations in Travel Modeling, Portland, OR, June 22-24.
- 34 Kraan, M., Mahmassani, H.S., and Huynh, N. (2000). Traveler Responses to Advanced Traveler
35 Information Systems for Shopping Trips: Interactive Survey Approach. In *Transportation Research*
36 *Record: Journal of the Transportation Research Board*, 1725, 116-123.
- 37 Levinson, D. (2003). The Value of Advanced Traveler Information Systems for Route Choice.
38 *Transportation Research C*, 11(1), 75-87.
- 39 Lin, D-Y., Eluru, N., Waller, S.T., and Bhat, C.R. (2008). Integration of Activity-Based Modeling and
40 Dynamic Traffic Assignment. In *Transportation Research Record, Journal of the Transportation*
41 *Research Board*, 2076, 52 - 61.
- 42 Liu, H. and Mahmassani, H.S. (1998). Dynamic Aspects of Commuter Decisions Under Advanced Traveler
43 Information Systems: Modeling Framework and Experimental Results. In *Transportation Research*
44 *Record: Journal of the Transportation Research Board*, 1645, 111-119.
- 45 Paz, A. and Peeta, S. (2008). Paradigms to Deploy a Behavior-Consistent Approach for Information-Based
46 Real-Time Traffic Routing. *Networks and Spatial Economics*, 9(2), pp. 217-241.
- 47 Pendyala, R.M., Konduri, K.C., Chiu, Y., Hickman, M., Noh, H., Waddell, P., Wang, L., You, D., and
48 Gardner, B. (2011). An Integrated Land Use-Transport Model System with Dynamic Time-

1 Dependent Activity-Travel Microsimulation. In *Transportation Research Record: Journal of the*
2 *Transportation Research Board*, in press.

3 Sundaram, S. (2002). Development of a Dynamic Traffic Assignment System for Short-term Planning
4 Applications. Master's thesis, Massachusetts Institute of Technology, Cambridge, MA.

5 Waddell, P.A., Liu, X., and Wang, L. (2008). UrbanSim: An Evolving Planning Support System for Evolving
6 Communities. In R.K. Brail (ed.) *Planning Support Systems for Cities and Regions*, Lincoln Institute
7 of Land Policy, Cambridge, MA, Chapter 6, pp. 103-138.

8 Yun, M., van Herick, D., and Mokhtarian, P.L. (2011). Nonwork Travel Behavior Changes During
9 Temporary Freeway Closure: The Fix I-5 Project in Sacramento, California. In *Transportation*
10 *Research Record: Journal of the Transportation Research Board*, 2231, 1-9.

11 Zhu, S., Levinson, D., Liu, H.X., and Harder, K. (2010). The Traffic and Behavior Effects of the I-35W
12 Mississippi River Bridge Collapse. *Transportation Research A*, 44(10), 771-784.

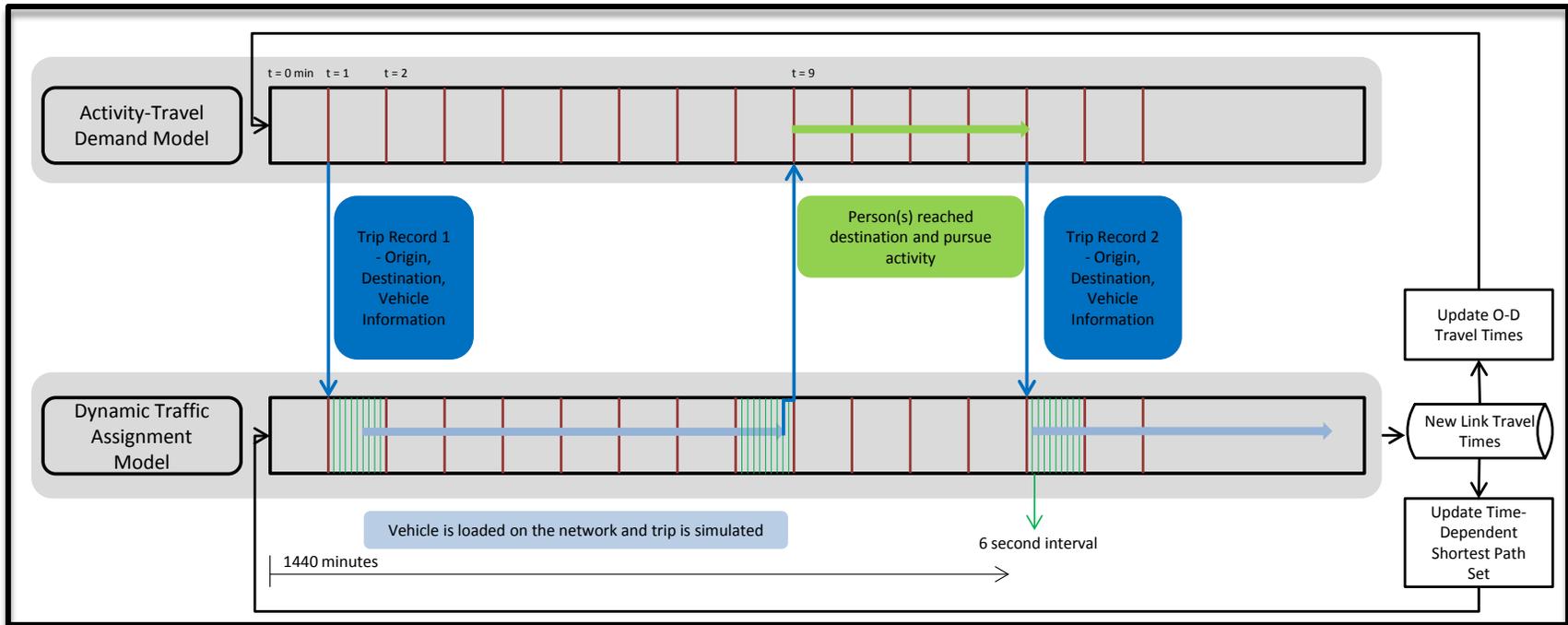


FIGURE 1. Network-Sensitive Activity-Travel Simulation Framework Incorporating an Event-Based Paradigm

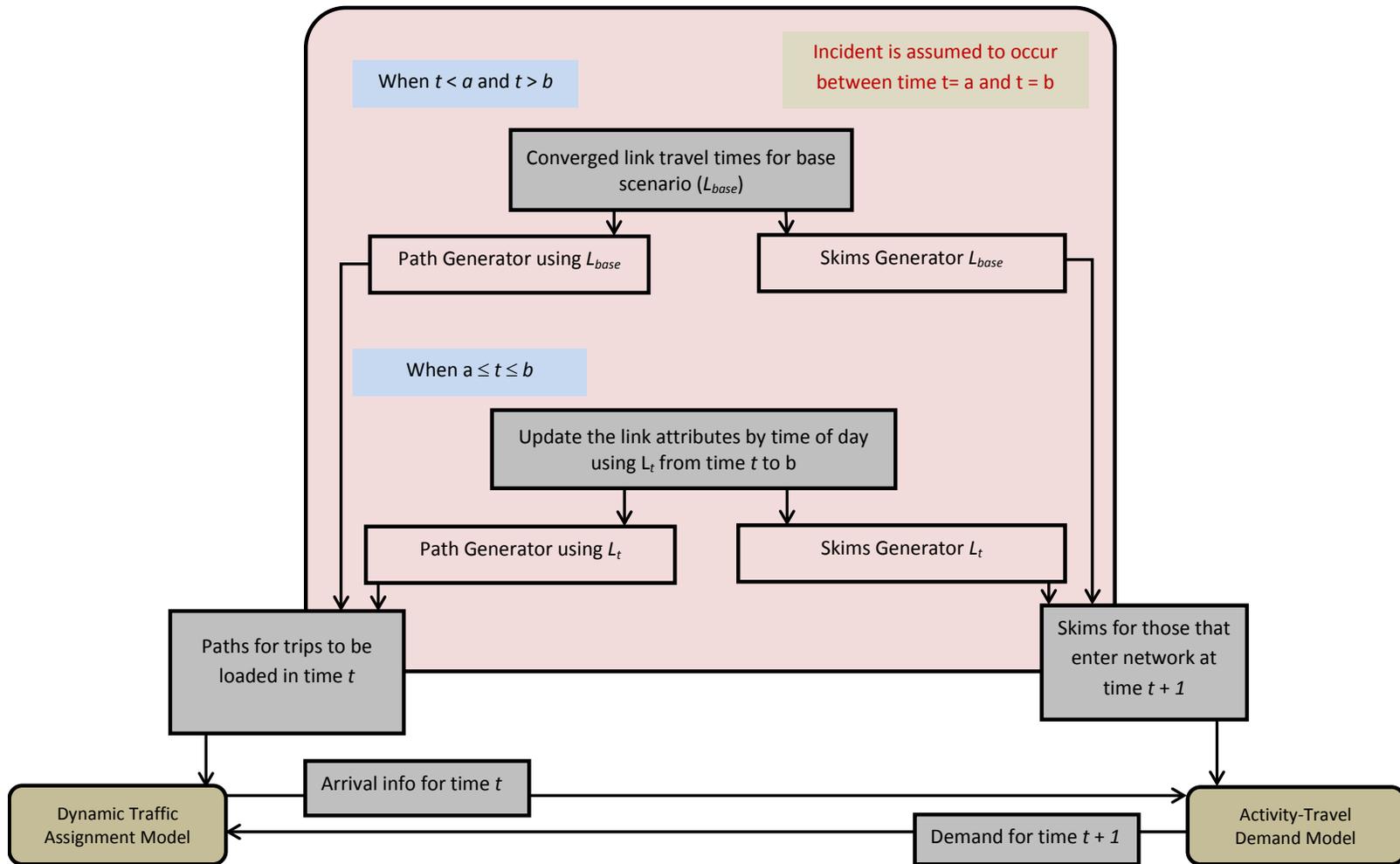


FIGURE 2. Modified SimTRAVEL Framework to Simulate Information Provision Strategies Under Network Disruptions

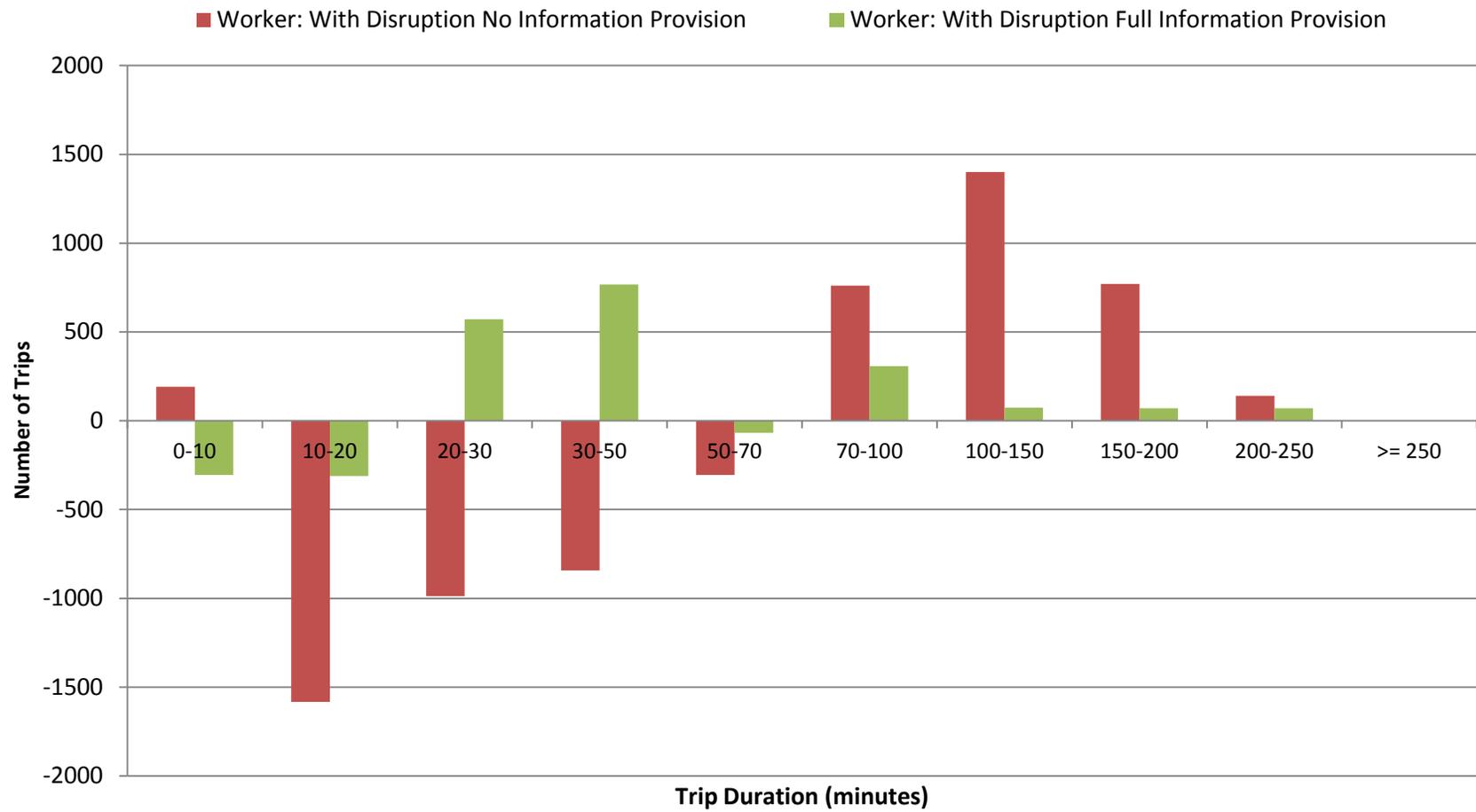


FIGURE 3. Difference in Trip Duration Distributions for Workers

TABLE 1. Difference in Daily Time Allocation for Trips and Activities Across All Individuals in Person-Days

	Worker		Non-worker	
	No Information Provision	Full Information Provision	No Information Provision	Full Information Provision
Number of people	150435	150435	187757	187757
<i>Time Spent on Activities</i>				
Home	-72	8	-71	-31
Work	-111	-28	0	0
School	4	4	-1	1
Maintenance	-21	-50	-62	-7
Discretionary	-15	-4	-8	13
Pick Up	1	1	2	2
Drop Off	1	1	8	2
OH-Other	-3	-3	9	5
<i>Total activity duration</i>	<i>-217</i>	<i>-71</i>	<i>-123</i>	<i>-15</i>
<i>Time Spent on Trips</i>				
Home	46	31	33	9
Work	110	23	0	0
School	2	0	2	0
Maintenance	21	7	22	-9
Discretionary	8	4	12	10
Pick Up	6	0	15	1
Drop Off	24	3	34	7
OH-Other	4	0	7	-3
<i>Total travel duration</i>	<i>221</i>	<i>69</i>	<i>124</i>	<i>16</i>