Applying Analysis Tools in Planning for Operations

Case Study #4 – Application of Microsimulation in Combination With Travel Demand Models
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More and more, transportation system operators are seeing the benefits of strengthening links between planning and operations. A critical element in improving transportation decision-making and the effectiveness of transportation systems related to operations and planning is through the use of analysis tools and methods. This brochure is one in a series of five intended to improve the way existing analysis tools are used to advance operational strategies in the planning process. The specific objective of developing this informational brochure series was to provide reference and resource materials that will help planners and operations professionals to use existing transportation planning and operations analysis tools and methods in a more systematic way to better analyze, evaluate, and report the benefits of needed investments in transportation operations.

The main problem of interfacing travel demand models with microsimulation models is that the demands produced by demand models are not as capacity constrained as they need to be for use in microsimulation models. The result is that the microsimulation model produces unrealistic facility performance estimates when it is given unrealistic calibration year and future year demands. The solution is to adjust the travel demand model demands to more realistic levels that reflect the physical limitations of the network. This brochure describes two case studies for applying simulation models in combination with travel demand models: 1) the traditional approach performing these adjustments outside of the travel demand model; and 2) an advanced approach that makes many of the demand adjustments within the demand model.
### SI* (MODERN METRIC) CONVERSION FACTORS

#### APPROXIMATE CONVERSIONS TO SI UNITS

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NOTE: Volumes greater than 1000 L shall be shown in m³

| **MASS** | | | | |
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| lb | pounds | 0.454 | kilograms | kg |
| T | short tons (2000 lb) | 0.907 | megagrams (or "metric ton") | Mg (or "T") |

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| or (°F-32)/1.8 |

| **ILLUMINATION** | | | | |
| fc | foot-candies | 10.76 | lux | lx |
| ft | foot-Lamberts | 3.426 | candela/m² | cd/m² |

| **FORCE and PRESSURE or STRESS** | | | | |
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| lbf/in² | poundforce per square inch | 6.89 | kilopascals | kPa |

### APPROXIMATE CONVERSIONS FROM SI UNITS

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| **TEMPERATURE (exact degrees)** | | | | |
| °C | Celsius | 1.8°C+32 | Fahrenheit | °F |

| **ILLUMINATION** | | | | |
| lx | lux | 0.0929 | foot-candies | fc |
| cd/m² | candela/m² | 0.2919 | foot-Lamberts | fl |

| **FORCE and PRESSURE or STRESS** | | | | |
| N | newtons | 0.225 | poundforce | lbf |
| kPa | kilopascals | 0.145 | poundforce per square inch | lbf/in² |

*SI is the symbol for the International System of Units. Appropriate rounding should be made to comply with Section 4 of ASTM E380. (Revised March 2003)*
Applying Analysis Tools in Planning for Operations

Case Study #4 – Application of Microsimulation in Combination With Travel Demand Models

More and more, transportation system operators are seeing the benefits of strengthening links between planning and operations. A critical element in improving transportation decision-making and the effectiveness of transportation systems related to operations and planning is through the use of analysis tools and methods. This brochure is one in a series of five intended to improve the way existing analysis tools are used to advance operational strategies in the planning process. The specific objective of developing this informational brochure series was to provide reference and resource materials that will help planners and operations professionals to use existing transportation planning and operations analysis tools and methods in a more systematic way to better analyze, evaluate, and report the benefits of needed investments in transportation operations.

The series of brochures includes an overview brochure and four case studies that provide practitioners with information on the feasibility of these practices and guidance on how they might implement similar processes in their own regions. The particular case studies were developed to illuminate how existing tools for operations could be utilized in innovative ways or combined with the capabilities of other tools to support operations planning.1 The types of tools considered when selecting the case studies included:

- Sketch planning tools;
- Travel demand forecasting models;
- Deterministic models;
- Traffic signal optimization tools;
- Simulation tools;
- Archived operations data;
- Operations-oriented performance measures/metrics; and
- Combinations of these tools and methods.

Additional information on these existing tool types is presented in the overview brochure to this series.

In selecting the case studies to highlight in this brochure series, a number of innovative analysis practices and tool applications were considered. Ultimately, four different case studies were selected from among many worthy candidates. Each of these case studies represents an innovative application of analysis tools to support operations planning.

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1 The use of the term “Tools” in this context is meant not only to include physical software and devoted analytical applications, but is also intended to encompass more basic analysis methods and procedures as well.
use of one or more of the tool types listed above. Figure 1 presents the topics of the case studies and maps them to the related tool. Although individual case studies were not developed for each tool category, this should not be considered as a measure of indictment of the ability of any tool type to be used in innovative ways to support operations planning – there simply weren’t project resources to identify and document all the innovative practices being used. Likewise, the selection of a particular case study representing a specific tool should not be construed as the only manner in which to apply the particular tool. Instead, the case studies represent a sampling of the many innovative ways planners and operations personnel are applying these tools currently.

Figure 1. Analytical Methods/Tools and Related Case Studies Developed Under this Project

**Case Study Introduction**

The main problem of interfacing travel demand models with microsimulation models is that the demands produced by demand models are not as capacity constrained as they need to be for use in microsimulation models. Demand models have a flexible capacity constraint; the traffic assigned to a facility during the analysis period can exceed its capacity by several orders of magnitude. Microsimulation models have a “storage space constrained capacity constraint”; the traffic assigned to a facility during the analysis period cannot exceed its capacity plus its ability to store the excess queues of vehicles. The result is that the microsimulation model produces unrealistic facility performance estimates when it is given unrealistic calibration year and future year demands.

The solution is to adjust the travel demand model demands to more realistic levels that reflect the physical limitations of the network (the flow capacity and the storage capacity). This section describes two
case studies for applying simulation models in combination with travel
demand models. The traditional approach described in the first case
study performs these adjustments outside of the travel demand model.
The second case study is an advanced approach that makes many of the
demand adjustments within the demand model.

**Case Study Objectives**

This case study documents the benefits as well as the pros and cons of
integrating travel demand forecasting models with microsimulation tools
for freeway operational studies.

**Statement of the Problem**

Current travel demand models and planning analyses may underestimate
the day-to-day benefits of operational improvements targeted at the
reduction of vehicular traffic congestion. Underestimation of the ongoing
benefits of relieving congestion results in a bias in the transportation
planning and programming process in favor of capital improvements,
which increase capacity over operational improvements.

Travel demand forecasting models are usually validated for regional
characteristics. For example, the Contra Costa County Transportation
Agency’s (CCTA) model, which was used for Case Study 4b in this
document, includes a detailed zone system throughout Contra Costa
County and the Alameda County portion of the Tri-Valley area. It was not
designed for corridor analysis or sensitivity analysis of various freeway
management strategies.

In general, travel demand forecasting models have the following
limitations:

- **Neglect vehicular lane changing behaviors.** Travel demand
  forecasting models usually do not account for reductions in
  actual roadway capacity due to weaving, merging, and diverging
  characteristics. Even though the link capacity can be adjusted in
  travel demand models, the adjustment (reduced link capacity)
  can only be applied throughout whole links with a fixed averaged
  number. In reality, these adjustments may not be appropriate for the
  traffic operations.

- **Neglect roadway characteristics: grades, curvatures
  (interchanges).** Travel demand models generally contain lesser level
  of detail of roadway characteristics such as grades, curvatures, etc.
  These characteristics usually influence traffic operations strongly,
  especially at freeway interchanges.

- **Neglect congested roadway conditions: peak hour spreading,
  traffic diversion.** By definition, in a travel demand model where
  travel demand (assignable trip table) is greater than the capacity of
  roadway systems, demand will still be assigned onto the network
  even though this results in links with volume to capacity ratios
  greater than 1.00. However, these v/c ratios which should represent
  queuing and potential bottlenecks, do not affect upstream links.
  In other words, in a travel demand model traffic cannot “back up”
  from an overcapacity link into an upstream link and vehicular route
  choices are not affected by the downstream congestion unless the
  congested link is part of the route.

Because of these limitations of travel demand forecasting models, it is
essential to apply microsimulation models for evaluating the full benefits
of freeway management strategies.
Case Study 4a - Traditional Application of Travel Demand Models and Microsimulation Models

This first case study involves a conventional application of a microsimulation model in combination with a travel demand model.

The travel demand model is used to estimate existing and future origin-destination (OD) demands for a freeway section. The calibration year OD was then adjusted to match the calibration year counts for the freeway. These calibration year OD adjustments were then carried forward to the future year forecasts and applied to the future year OD trip tables produced by the demand model. The microsimulation model is applied using the adjusted calibration year and future year OD tables.

The freeway performance is estimated exclusively using the microsimulation model. The demand model performance predictions are not used.

Project Description

The goal of the Alameda County (California) Central Freeway Study was to prioritize a funding sequence among various combinations of all potential freeway improvement projects in the jurisdiction.

The Alameda County Central Freeway Study evaluated 10 miles of the I-880 freeway from the Davis Street (SR 112) interchange to Whipple Road and five miles of I-238/I-580 freeway from 164th Street to East Castro Valley Boulevard. Figure 2 shows the freeway network in the study area. The Alameda County Congestion Management Agency’s (ACCMA) official travel demand model (developed in Cube) was used to forecast future travel demand in the corridor, including ramp and mainline volumes on the freeways. Paramics microsimulation software was selected for producing measures of effectiveness (MOE) results of freeway operations for each alternative. Traffic conditions of morning and afternoon (AM and PM) peak hours were evaluated.

Surface streets were not modeled in the microsimulation model. Surface streets were included only in the ACCMA Cube travel demand model. Two freeway interchanges and 20 ramp junctions (on-ramps and off-ramps) were included in the microsimulation network.

Methodology

The approach used is outlined below:

1. Forecast base-and-future year regional travel demand models,
2. Extract subarea networks,
3. Produce origin-destination (OD) matrices,
4. Apply OD matrices (trip tables) to microsimulation models,

5. Validate and calibrate the base year microsimulation models, and

6. Create the future base and project microsimulation models.

Figure 2. Alameda County Central Freeway Study Area

Figure 3 shows the flowchart of this traditional method. The details of these steps are explained in the following paragraphs:

1. **Prepare the Base-year OD Matrices**

The base-year travel demand model for this case study contained the land use data for 2005. The model’s loaded 2005 network (with traffic assignments) was used to generate the subarea network. The Cube software has a function to generate OD matrices (AM and PM peak-hour trip tables) based on a subarea network. Thus, analysts simply “cut” the large regional network into a smaller subset area and then run the scripts. For this case study, this process consolidated 3,000 zones (the regional model) into 43 zones of the subarea network.
2. Import the Base-year OD Matrices into a Microsimulation Model

Each microsimulation software package handles OD matrices differently. In Paramics, multiple OD matrices can be loaded onto the same network file. For example, one network file can contain two OD matrices: AM and PM peak hour. The zone numbers in Paramics network should be consistent with the subarea demand model’s zone numbers.

Once the base year OD matrix is imported into the microsimulation model, analysts start the validation and calibration processes.

3. Validate and Calibrate the Base Year Microsimulation Model

In this study area, based on field data at certain locations on westbound freeway I-580, the maximum capacity of the freeway was roughly 2,000 cars per lane. However, the demand model loaded the network with almost 2,500 cars per lane. During the initial microsimulation runs, the traffic flow of the westbound I-580 was completely shutdown near the diverging point of the freeway I-238, so the downstream freeway segments received very few traffic volumes. Since this level of demand cannot enter the network during the single peak hour, the analysts needed to reduce the assigned demand by adjusting the base year OD matrices to remove the unrealistic bottleneck. As a result, the downstream freeway segments received reasonable traffic volumes and the measure of effectiveness was able to be gathered and presented properly.

When the assigned traffic flow was reduced to something more realistic on freeway I-580, some bottlenecks were found on both northbound
and southbound directions of freeway I-880. Analysts again checked the roadway geometry and adjusted the OD matrices in a second round. These trial-and-error processes require tremendous effort when the study area is relatively large. Calibrating and validating the microsimulation model for the base year according to the FHWA’s microsimulation guidelines consumed a large portion of the budget.

4. Build Future Base and Project Scenario Microsimulation Models

The year 2025 future ACCMA models (with traffic assignments) were used to create future year trip tables. Analysts extracted the regional models into the smaller subarea. Cube scripts were applied to the subarea and generated the future trip tables.

The same percentages of volume adjustments (as were applied to the calibration year trip table) were applied to the future trip tables. In this case study, analysts used Microsoft Excel software to document all adjustments of the base-year trip tables. Thus, when future trip tables were created by the ACCMA model, analysts were able to simply copy and paste the adjustments (equations) to the future no-project and future project trip tables.

After the base year microsimulation model was complete, analysts copied the model and saved it as the future “no-project” network. New geometry was added on the network allowing for the incorporation or uploading of future year OD matrices in the models. Simulation runs were performed and the models were also checked. Therefore, project scenario networks were created based on the future base microsimulation networks. All estimates of measures of effectiveness for each project scenario were gathered and compared with the future “no-project” simulation models.

Pros of the Traditional Approach

This case study demonstrated the benefit of combining a simulation model with a demand model to evaluate the benefits of a freeway improvement project.

The simulation model results showed that some systemwide benefits of certain project scenarios were off-set by the increased volumes. For example, one of the project scenarios was to modify the one-lane on-ramp to a two-lane on-ramp at the merging area from the I-238 freeway to the southbound I-880. In existing conditions, this capacity constraint (one-lane on-ramp) caused the queue on the westbound I-238 and sometimes even spilled back onto the I-580 freeway. When the ramp capacity increased from one-lane to two-lanes, it brought roughly one thousand more vehicles onto the southbound I-880. These increased volumes resulted in more delay of the traffic flow on southbound I-880. Thus, the overall travel time saving was less than the agency’s presumption.

At the end of this project, the benefits of applying microsimulation in combination with travel demand models were shown and helped the agency to prioritize the funding sequence of all project scenarios.

Cons of the Traditional Approach

The traditional approach (adjusting the demand outside of the demand model) is feasible to perform manually (with the assistance of a spreadsheet) for small microsimulation study areas employing no more than 50 origin and destination zones. This approach becomes too laborious for larger study areas. Larger microsimulation study areas would require greater automation of the post-demand model adjustment process.

Besides the physical limits on the ability of the analyst to manually adjust large OD trip tables, there is also the theoretical concern that the demand adjustments are being made on an “ad-hoc” basis, without taking advantage of the behavioral models already incorporated at great expense in the demand model. The analyst simply reduces the calibration year demands produced by the demand model to match existing counts and then assumes that the same errors are also present in the future year forecasts produced by the demand model. This assumption does not take into account changes in the future network (more transit service for example) or the implications of reducing demand in one corridor on the operation of nearby corridors.

Case Study 4b - Innovative Integrated Applications of Travel Demand Models and Microsimulation Models

The goal of this more innovative approach to combining travel demand models with microsimulation is to reflect the effects of downstream weaving and queuing on upstream locations (the output of the microsimulation model) within the travel demand model itself. This procedure that includes validation of an estimated base year trip matrix in the travel demand model resulting in acceptable congested speeds and queues in the base-year simulation model as well as the development of a growth matrix, was used for the second case study described below.

Project Description

The Tri-Valley area is nestled between major job centers in Silicon Valley and affordable housing supply areas east of the San Francisco Bay Area (San Joaquin Valley) in California. The cities in the Tri-Valley area (Dublin, Pleasanton, and Livermore) are also undergoing massive development in housing and employment. The primary transportation corridors serving travel to and through this area are already over-capacity today for several hours during the morning and afternoon peak periods. Significant volumes of traffic divert from the freeways to parallel local streets. The Alameda County Congestion Management Agency (ACCMA) initiated the Triangle Study to evaluate and develop a near-term and long-range plan for sequencing improvements for practical traffic relief on the Tri-Valley freeways (I-580, I-680 and SR 84) in a cost effective manner consistent with the transportation needs in the area.

The regional travel demand model used for this case study was the Contra Costa Transportation Authority’s (CCTA) Decennial model. This regional model has 1,454 traffic analysis zones covering the entire Bay Area (Figure 4) with more detail in Contra Costa County and the Tri-Valley area. Full model runs were performed for the existing and future base years and a subarea highway network and associated trip tables were extracted (Figure 5).
Methodology

This example of an innovative approach of combining travel demand models with microsimulation models was performed as follows:

1. Refinement of the full regional model network and traffic analysis zones to reflect and support the Tri-Valley cities’ level of detail;

2. Development of each Tri-Valley city’s land use projections for base-year and build-out scenarios (different from Association of Bay Area Governments (ABAG) 2030);

3. Full application of the regional model for base-year and build-out scenarios (development of the peak four-hour assignments and trip tables);

4. Extraction of the subarea networks and peak period (four-hour) trip tables for base-year and build-out scenarios;
5. Application of the resultant base-year trip table matrices onto the existing conditions’ microsimulation network;

6. Feedback from the traffic microsimulation model into the travel demand forecasting base-year model to reflect travel diversion caused by queuing and weaving;

7. Validation of the subarea demand models with implementation of a peak hour spreading algorithm using matrix estimation to ensure that the travel demand model reflects at-capacity conditions based on existing traffic counts and limits on link capacity due to queuing and bottlenecks;

8. Development of future base build–out trip tables using forecast growth in peak-period traffic from existing to build-out conditions and implementation of the peak-hour spreading algorithm using matrix estimation to ensure that the travel demand model reflects at-capacity conditions based on future regional capacity constraints;

9. Application of the resultant future base build–out trip table matrices onto the build-out conditions’ microsimulation network;

10. Feedback from the future base traffic microsimulation model into the future base travel demand forecasting model to reflect travel diversion caused by queuing and weaving; and

11. Final export of the peak hour at capacity demand trip tables from the travel demand models into the microsimulation models.

Of these steps, the most significant difference from a traditional approach is the implementation of a peak spreading algorithm and the iterative feedback between the travel demand model and the microsimulation model.

Key modeling procedures are described below.

**Create Subarea from Regional Travel Demand Model – Existing Conditions**

The first step in the process was updating the regional model by adding network detail and splitting traffic analysis zones (TAZ) to allow for analyses of build out of the local city general plans and to reflect local access to the highway network. From the regional model, a subarea extraction process was used to create subarea networks and peak period trip tables. The full four-hour a.m. and four-hour p.m. traffic assignments were used to create the subarea networks to ensure that the full demand would be included. The subarea model had 600 traffic analysis zones.

**Adjust Subarea Demand Model (Validation)**

Matrix estimation (ME) was used to create one-hour subarea trip tables from the four-hour subarea traffic assignments using known supply (capacity) and demand (counts and cut-through surveys) assumptions. The latest peak one-hour traffic counts were used to validate the base year demand model as well as review of cut-through traffic. In essence, the base-year peak trip matrix was factored to better replicate observed traffic counts, cut-through travel patterns, and most importantly regional capacity constraints. This procedure is outlined in Chapter Eight of the report, NCHRP Publication 255. Cube software and its companion program Analyst (matrix estimator) were used for this project.
Balanced traffic volumes between intersections are critical for running any matrix estimator relying on traffic counts as a seed into the process. If traffic volumes are not balanced on the freeway and arterial corridors, the process cannot reasonably find optimal solutions which balance. Ideally, counts should be balanced before the matrix estimator process is run. The volume balancing function in Synchro software is a useful tool to perform the volume checking. Figure 6 shows the locations of available traffic counts for the Triangle Study.

![Traffic Count Locations](image)

The subarea model was validated to establish criteria including comparisons of model data to VMT from the Caltrans Highway Performance Monitoring System (HPMS), total volumes and percent root mean square error (RMSE) by facility types and volume groups, traffic counts across screen lines, and the percentage of links falling within the FHWA validation curve.

In addition to validating the travel demand model, the estimated demands were fed to the base-year simulation model to ensure that simulated congested speeds and queues were reasonable.

All members of the Technical Advisory Committee, including Caltrans and the participating local jurisdictional technical staff, were privy to the validation process.

**Apply the Adjustments of Existing Travel Demand to Future Travel Demand Model (Forecast)**

After the “fitted” at-capacity vehicle trip matrices were estimated for each time period (AM and PM peak periods), the increment of estimated growth between current and future conditions was calculated directly from the demand model and then added to the adjusted base year trip matrices. This process allows for the interplay of future growth while using a starting trip table which more appropriately represents existing at-capacity traffic patterns. The increments were added to the “unadjusted/original” forecasted models in appropriate time periods (AM and PM peaks) to produce adjusted model forecasts.
Adjusted Forecast Model = Future Base-year Forecast – Original Base-year Model + Adjusted Base Model

The adjusted model forecasts were estimated in this manner for each alternative to ensure consistent comparisons of MOEs between project alternatives.

**MOE Feedback and Integration**

After the existing and future base-year travel demand model trip tables were validated, they were imported to microsimulation models which allow for the analysis of reduced capacity due to merging, weaving, and queuing. In this case study, CORSIM software was used to evaluate traffic operations on the individual vehicle level.

Vehicle queuing and throughput information (served vehicles) can be fed back to the travel demand model. For example, if the freeway congestion is caused by downstream merging/diverging traffic bottlenecks, the constraint of this freeway throughput is not correctly represented in travel demand models. Thus, the delay calculated by microsimulation models can be fed back to the travel demand model to have more precise analysis results.

In this case study, delay information from the microsimulation model (in the form of reduced capacity) was fed back to the demand model. Analysts reran the traffic assignment of the travel demand models to obtain a modified trip table matrix which reflects additional rerouting of traffic based on effects of queuing and bottlenecks. This “second round” modified matrix was used by the microsimulation models for validating and calibrating the microsimulation networks.

This feedback loop was applied only once in this case study since the resulting queuing and congested speeds in the simulation model appeared reasonable but, depending upon the number of alternative routes and level of congestion, could be applied iteratively. Users may refer to the ITS Benefits Assessment Framework Study for more information regarding appropriate linkages between models.
Figure 7 shows the flowchart of the innovative approach applied in this second case study.

**Pros and Cons of the Innovative Approach**

The approach applied in this second case study takes into account known information about supply constraints (peak spreading) and travel demand patterns (cut-through traffic) as well as the effects of queuing and bottlenecks (microsimulation) on route diversion. This requires the iterative feedback between the travel demand model and the microsimulation model. While this “extra” step requires a level of effort, there is usually already a correspondence between the two models (since information must go from the travel demand model to the simulation model), which can be used to develop a correspondence in the “other” direction.
Also, as stated above in the discussion about Case Study 4a, if the trip tables that are fed to the microsimulation model do not take into account some level of peak hour spreading, the microsimulation models are difficult to calibrate and validate to existing conditions. This is particularly time-consuming and thus expensive for large-scale studies. Future levels of congestion simply exasperate the process. On the contrary, if analysts are not familiar with the appropriate application of a matrix estimation process in demand modeling, this procedure may take a significant amount of time. Many software packages provide this capability but, since there are many options in matrix estimating, engineering knowledge and judgment on what is happening in the study area are integral to making correct input adjustments and constraints such as setting confidence levels on origin-destination pairs, trip ends by traffic analysis zones, and the traffic counts.

**Case Study Outcomes**

While travel demand models reasonably forecast travel demand patterns which reflect a certain level of route diversion due to capacity constraints, they often fail when analysts assign trip tables representing high demands resulting in significant over-capacity conditions. By nature, a travel demand model (a macroscopic tool) will assign total volumes, regardless of whether the highway network supply can support it. Even though microsimulation models (a microscopic tool) will not allow the over assignment of travel demand (these vehicles are simply counted as “un-served”) this does not solve the problem when trying to output MOEs. In reality, the travel demand will “spread” to the shoulders and a certain amount of route diversion will occur. This is the goal of mesoscopic modeling.

Several software packages are developing the bridge between the macroscopic and microscopic (travel demand forecasting and microsimulation). Hopefully in the near future, analysts could have a more seamless integrated process between the two. In the mean time, some mesoscopic simulation tools can quantify impacts of upstream traffic congestion and measure queuing at intersections and merge points in a network. These tools integrate the feeding back of reductions in capacity information to the travel demand process. The innovative approach described herein attempts to apply this process without the availability of a reasonable mesoscopic tool.

The recommended approach to applying this process is to direct project resources (time and budget) to the validation of the future base scenario. The stakeholders will need to accept the results of the future base scenario so that the differences between alternatives can be used in the decision-making process. While validation of the models to base conditions is important, it is necessary to ensure that the sensitivity of the models to input growth assumptions is also validated. Then the model can be used to more reliably identify the differences in future alternatives.