



An interactive, visual computing
environment for the Portland Metro
activity-based travel forecasting
model

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Portland Metro ABM

DASH

- Dynamic Activity Simulator for Households
- Jon Gliebe, RSG

Emme

- Traffic, transit network models
- INRO

INRO is collaborating with Portland Metro and RSG to develop a coherent user interface, operating environment and reporting/query interface for a new activity-based model. This has been a chance to introduce some exciting new technologies for running models as well as for visualization and analysis.

Project goals (1/2)

“An efficient, well-structured interface to the model”

“Intuitive model operation”

“Logic that checks quality of input”

“Minimize potential for human error”

- Model spans different tools
- Complex procedures, many parameters
- Reproducibility

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Project goals include a usable, reproducible system in which to perform model specification, model production and model audit.

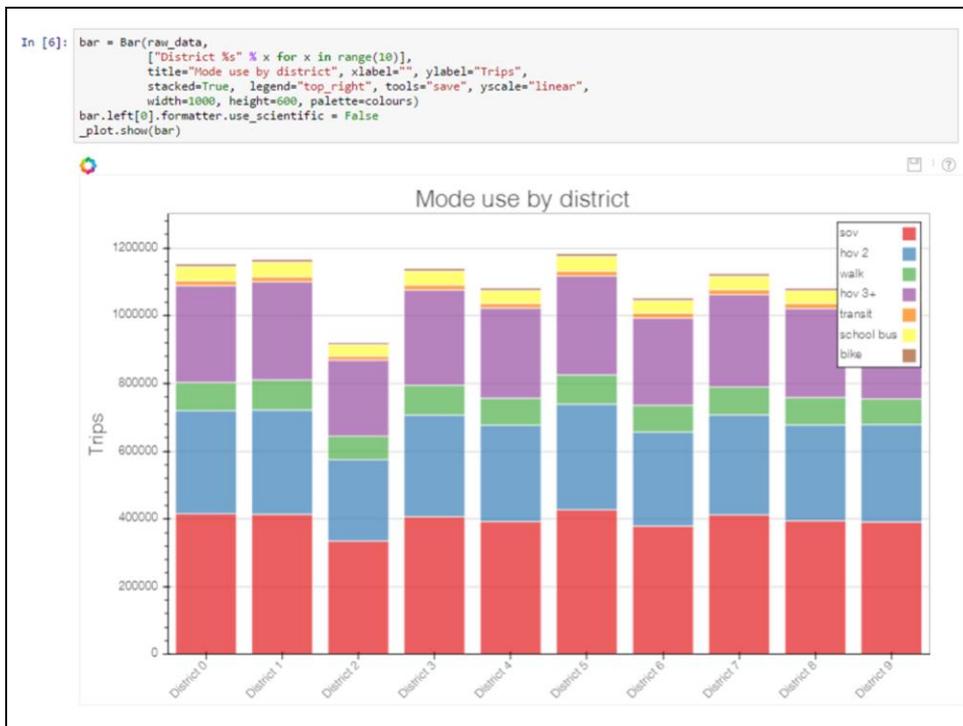
Project goals (2/2)

“Report generator” / “Query” /
“Visualization”

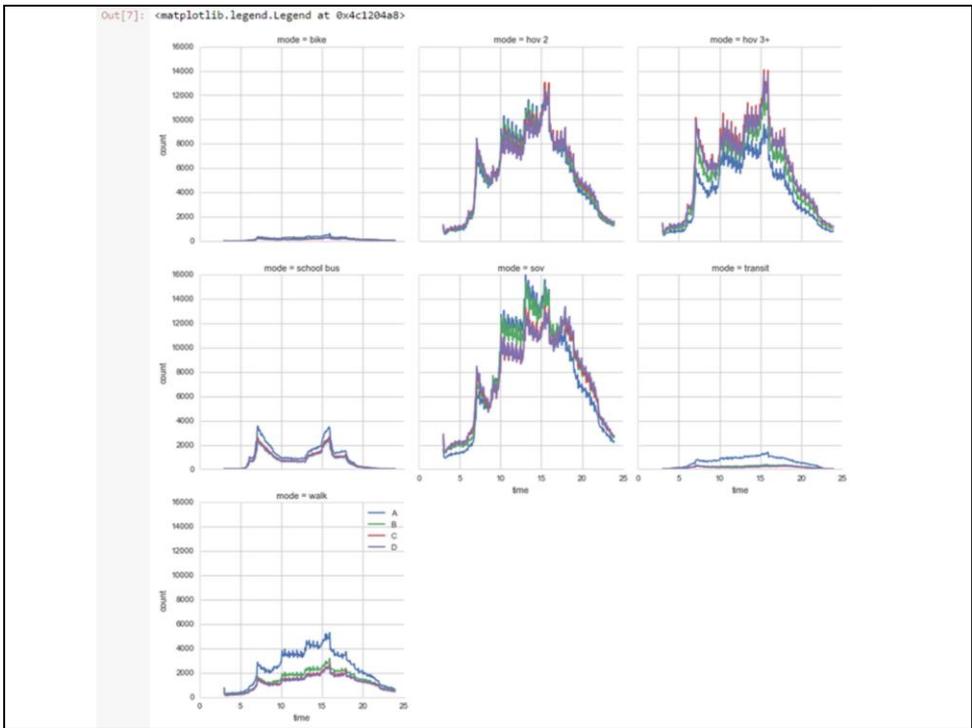
- Big spatial data
- Discovery, reasoning, communicating still hard

 INRO

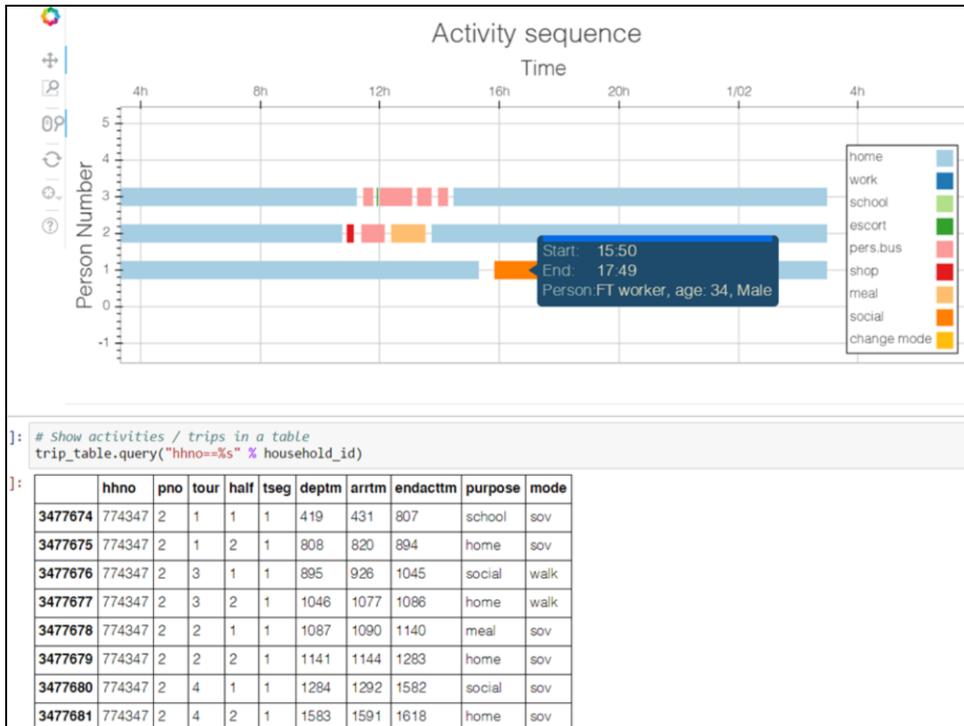
Reporting, query and visualization of ABM output is also a key project goal; one with fundamental challenges for which we propose a solution here.



The IPython Notebook is an interactive computing system which mixes text, code execution, math, graphics and other media in literate, reproducible computational documents. The project is open-source and intended to facilitate computational workflows across individual exploration, collaboration, production-scale execution and communication, making it a natural technology choice. Using popular Python libraries associated under the SciPy and PyData banners provides scientific and plotting libraries for analytics, like this chart showing mode use by district...



...and this mode use by time of day and income panel...



...and these detailed household activity sequence diagrams. Again, all the charts illustrated thus far are developed using open-source tools.

Mapping and exploring ABM data

- Disaggregate data is noisy and unwieldy
- But aggregations are lossy
- And can be time consuming
- Can we do better visual information seeking?
 - Overview first, zoom/filter, details-on-demand

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But mapping, animation and visual interaction with ABM output, which consists of tens of millions of trip trajectories and activities for even medium-sized cities, breaks down using traditional tools. We can do better.

Considerations

- 3.6 million people / 1.6 million parcels
- 11 million daily activities
- 14 million daily trips

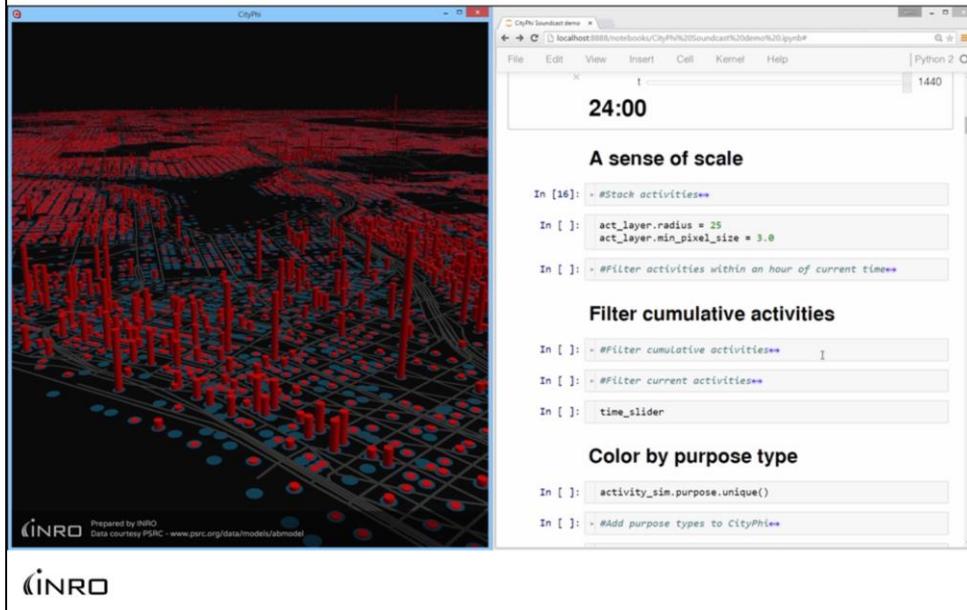
- Load time (<1 min)
- Query and filter time (<1 sec)
- Frame rate (target > 60 fps)
- RAM (up to 6GB)
- Mainstream hardware

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In order to properly support visual analytics with ABM output data, we need to be able to query and filter disaggregate data quickly, and explore space and time interactively. We'd like this to happen on reasonable hardware, too.

DASH ABM data for Portland is still being delivered, so instead we show examples from the metropolitan Seattle area, graciously provided by PSRC (<http://www.psrc.org/data/models/abmodel/>)

Activity analytics

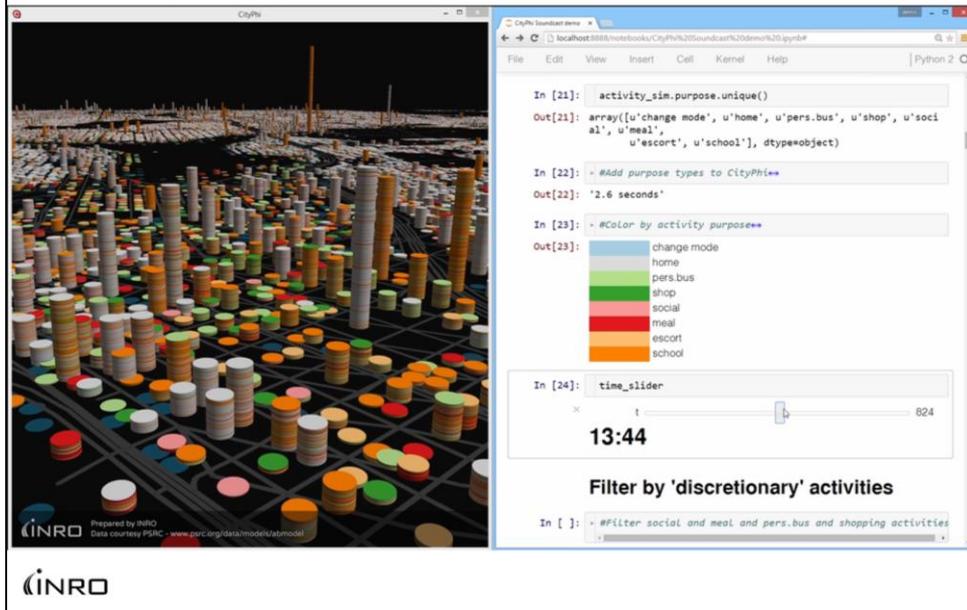


The right-hand side in the ensuing video demonstrations shows the IPython Notebook, discussed briefly earlier.

The left-hand side is a new INRO-developed framework for visual analytics of large-scale spatial and mobility data. This works together in core memory to meet the computational requirements laid out on the last slide, and to facilitate visual analytics on ABM data.

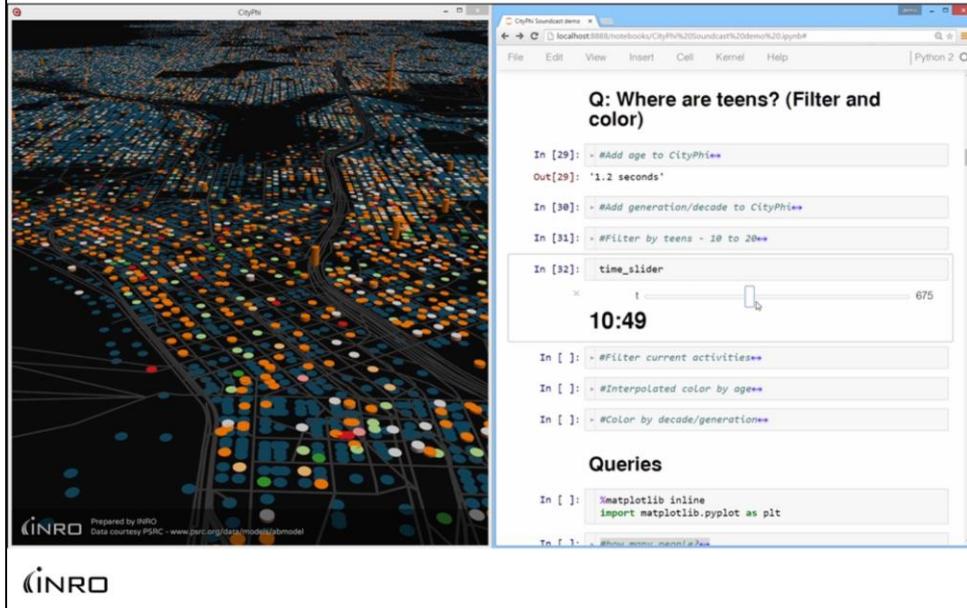
This image shows disaggregate activities displayed vertically at parcels and animated by time. The ensuing video stills provide additional examples of visual analytics.

Activity analytics



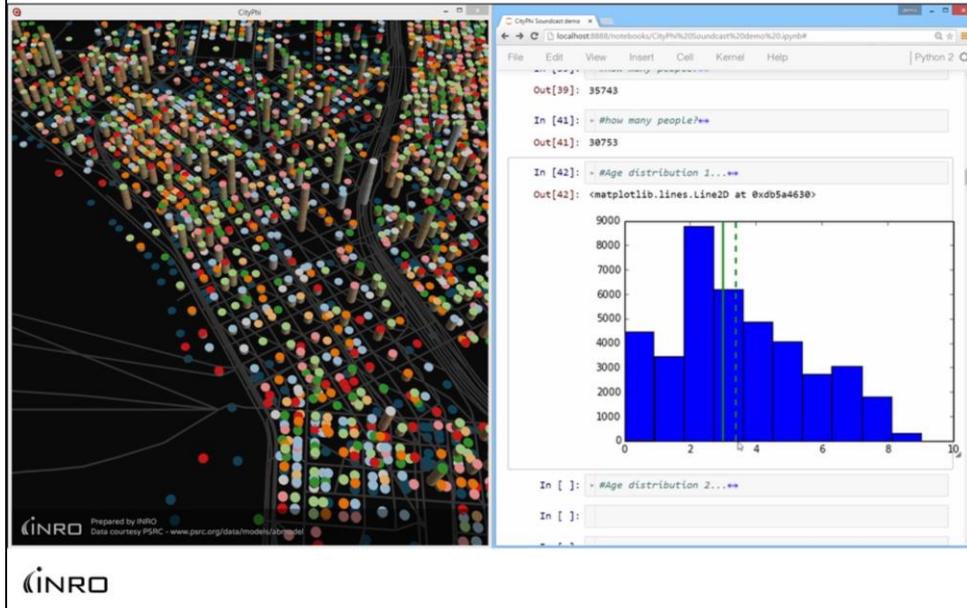
This image shows activities colored by purpose at 1:44pm. Working in core memory with the notebook, it takes 2.6 seconds to make activity purposes available for visual analytics, facilitating 'discussion time' visual analytics.

Activity analytics



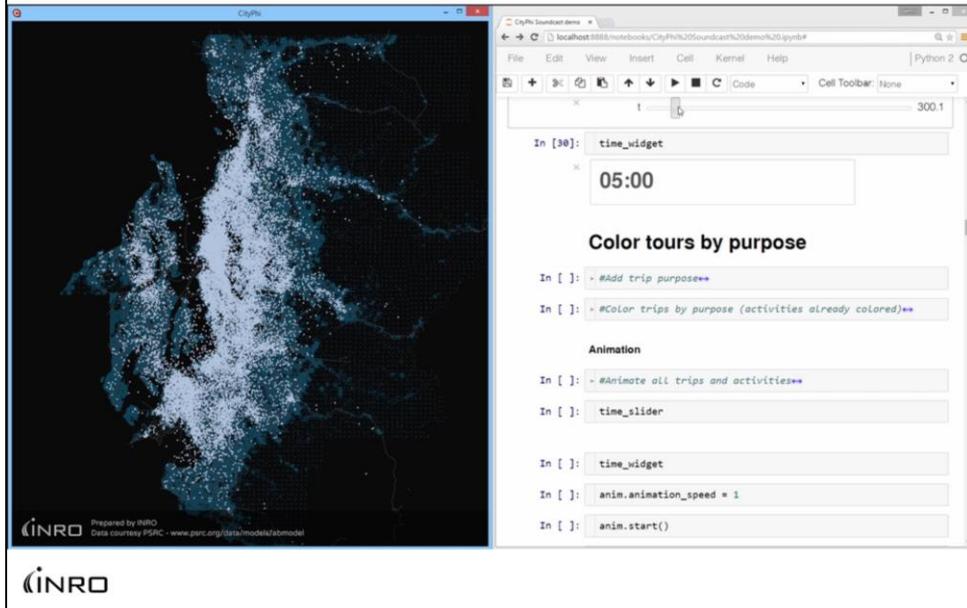
Uploading age information in 1.2 seconds and filtering by teens (10-20 year olds). At 10:49am many of them are in school (orange).

Activity analytics



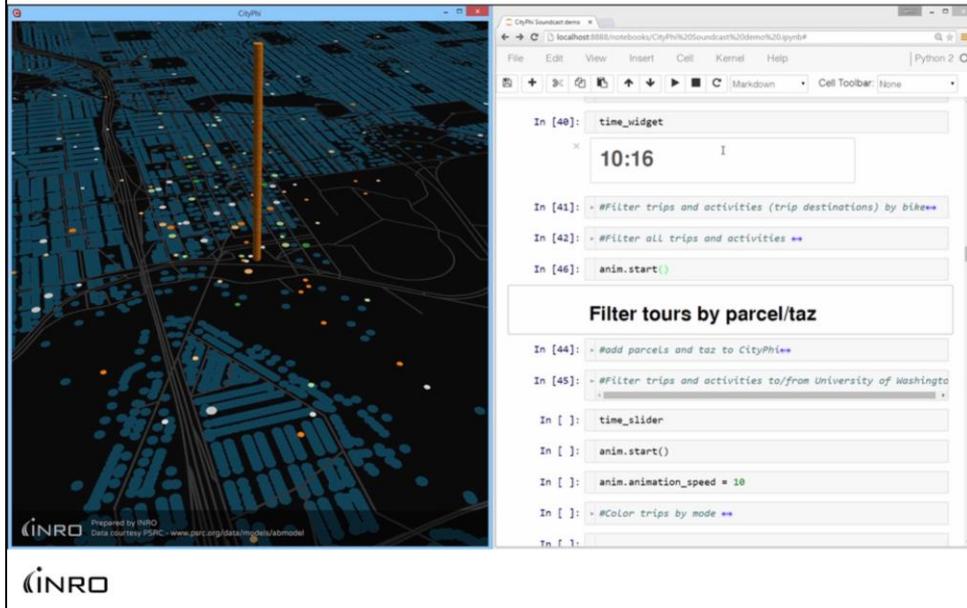
Keeping the data entirely disaggregate means that it can be queried. Here, we produce an age distribution of over 30000 people in downtown Seattle in the early afternoon, interactively. This can be easily compared with other regions or time periods of interest.

Tour tracing



Tour tracing shows as-the-crow-flies animations of mobility, from a fully synthesized set of >10,000,000 daily trips. (Note: work trips are excluded from this data set due to data privacy requirements)

Tour tracing



The screenshot displays a Jupyter Notebook environment. On the left, a 3D city model is shown with a vertical orange line indicating a specific location. The notebook interface on the right contains several code cells:

```
In [40]: time_widget
x
10:16

In [41]: - #filter trips and activities (trip destinations) by bike**
In [42]: - #filter all trips and activities **
In [46]: anim.start()

Filter tours by parcel/taz

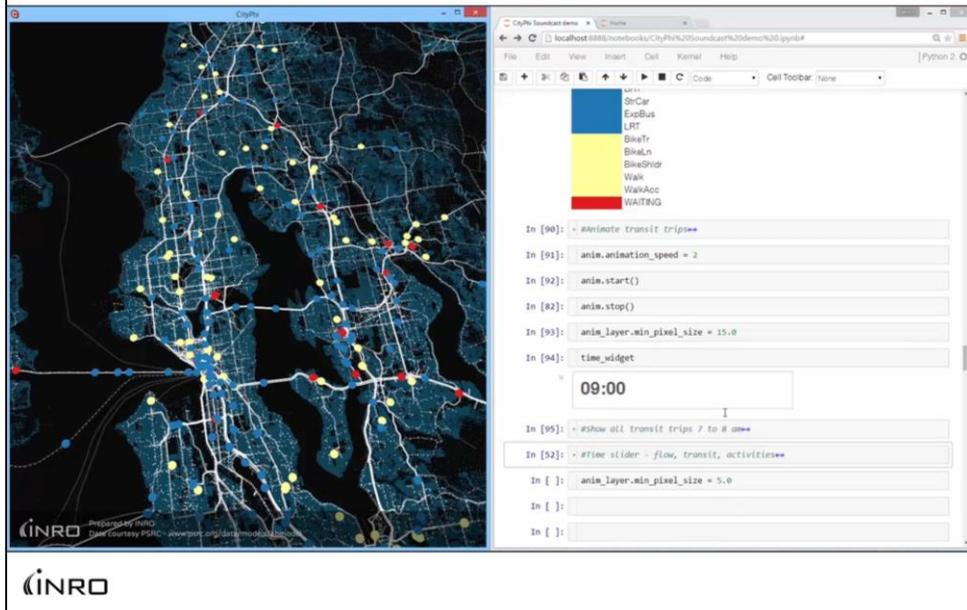
In [44]: - #add parcels and taz to CityPhi**
In [45]: - #filter trips and activities to/from University of Washingto
.

In [ ]: time_slider
In [ ]: anim.start()
In [ ]: anim.animation_speed = 10
In [ ]: - #Color trips by mode **
In [ ]:
```

The INRO logo is visible in the bottom left corner of the notebook interface.

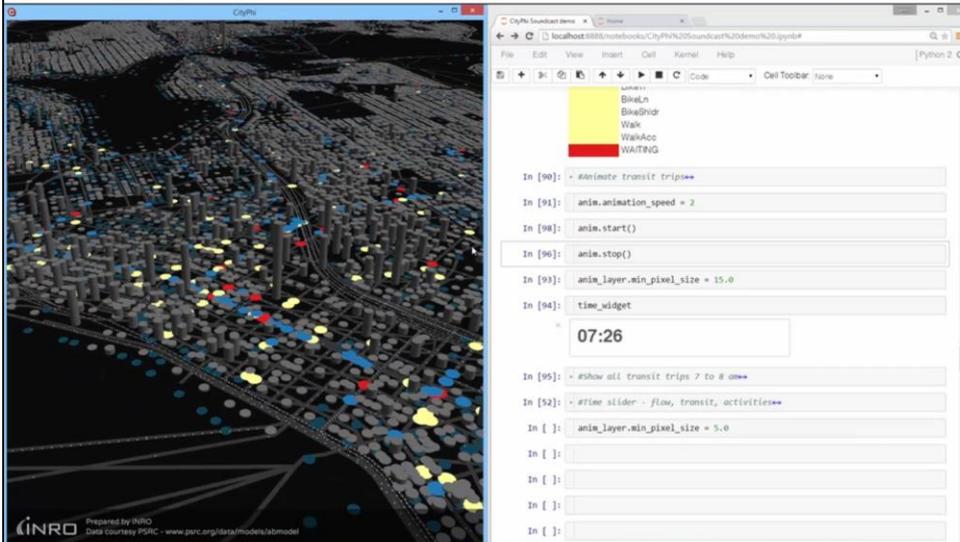
Activities and tours can also be filtered by parcel or TAZ, for instance, to show the rate of arrival at the university in the AM, and the corresponding parcels from which students and faculty arrive.

Trip tracing



Traffic flow and transit trips can also be animated, filtered, colored and queried. Here paths from a morning transit assignment are used to visualize the in-vehicle, walk/access/egress and waiting portions of discrete transit trips.

Trip tracing



The screenshot displays a Jupyter Notebook environment. On the left, a 3D city model is shown with various colored dots (yellow, blue, red) representing different activities or trip segments. The right side of the notebook shows a code editor with the following content:

```
Layers:  
BikeAcc  
BikeShare  
Walk  
WalkAcc  
WAITING
```

```
In [90]: - animate transit trips==  
In [91]: anim.animation_speed = 2  
In [98]: anim.start()  
In [96]: anim.stop()  
In [93]: anim_layer.min_pixel_size = 15.0  
In [94]: time_widget  
= 07:26  
In [95]: - show all transit trips > to # am==  
In [52]: - #time slider - flow, transit, activities==  
In [ ]: anim_layer.min_pixel_size = 5.0  
In [ ]:  
In [ ]:  
In [ ]:
```

INRO Prepared by INRO
Data courtesy of INRO - www.psrc.org/data/models/abmodel

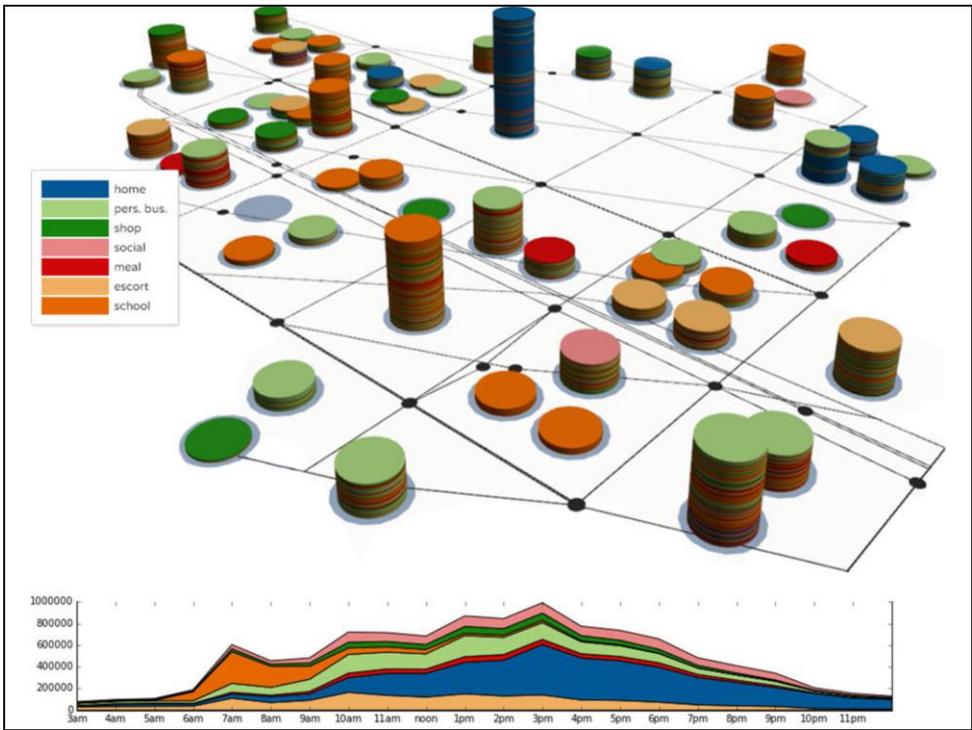


Interactive computing stack

- IPython Notebooks
 - DASH
 - ABM
 - Emme
 - Network models, network maps, charts
 - CityPhi
 - Visual analytics for large-scale spatial data
 - Python libs (Pandas, Numpy, Matplotlib...)
 - Data manipulation, charting, plotting



The IPython Notebook aims to solve interactive, reproducible computation, and together with Python and many popular open-source libraries is quickly becoming a standard in scientific computing. We are excited about applications to travel demand and transportation forecasting, with both PSRC and Portland Metro investing in corresponding staff skills. DASH, Emme and CityPhi will participate in this framework at Portland Metro.





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