



Modelling electric vehicle demand in London using the DCE platform

Dr Koen H. van Dam

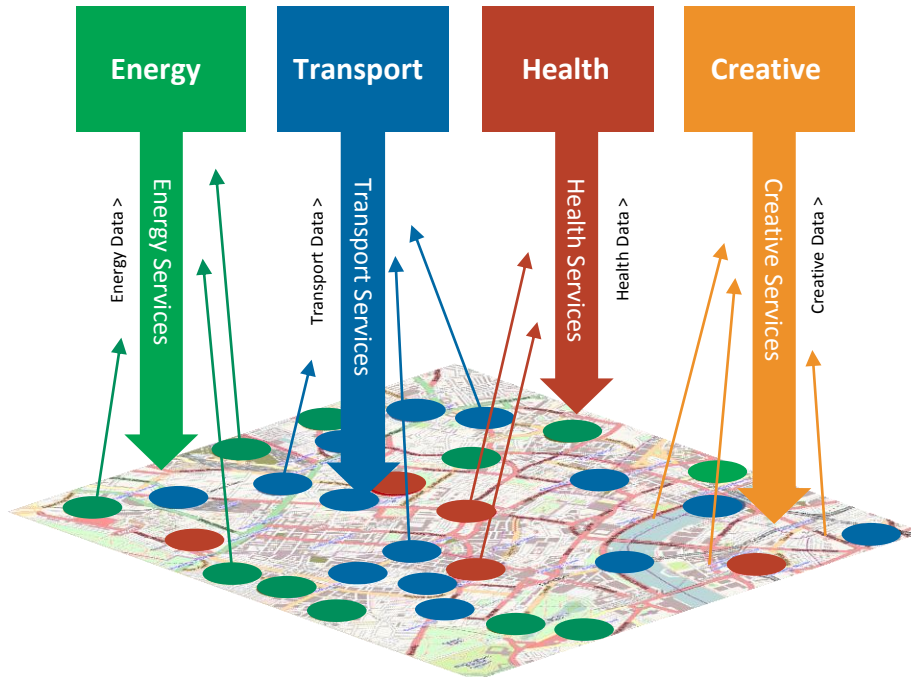
Systems-NET Webinar series
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Digital City Exchange

- A “smart city” is a connected city: efficient use of resources through interaction and integration
- Requires better understanding of the complexity of cities and urban living
- This is not a new idea, but maybe it can now happen:
 - Networks everywhere
 - Large-scale modelling
 - Pervasive sensing
 - Internet of things
 - Cloud computing
 - Etc...
- Connecting physical and digital

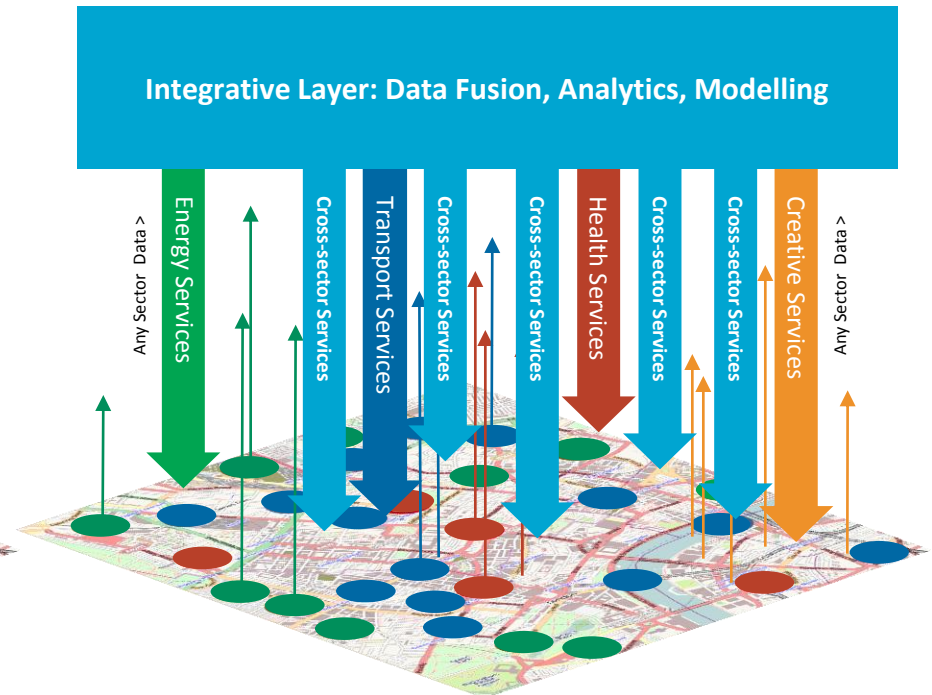
Conventional Data to Services Routes

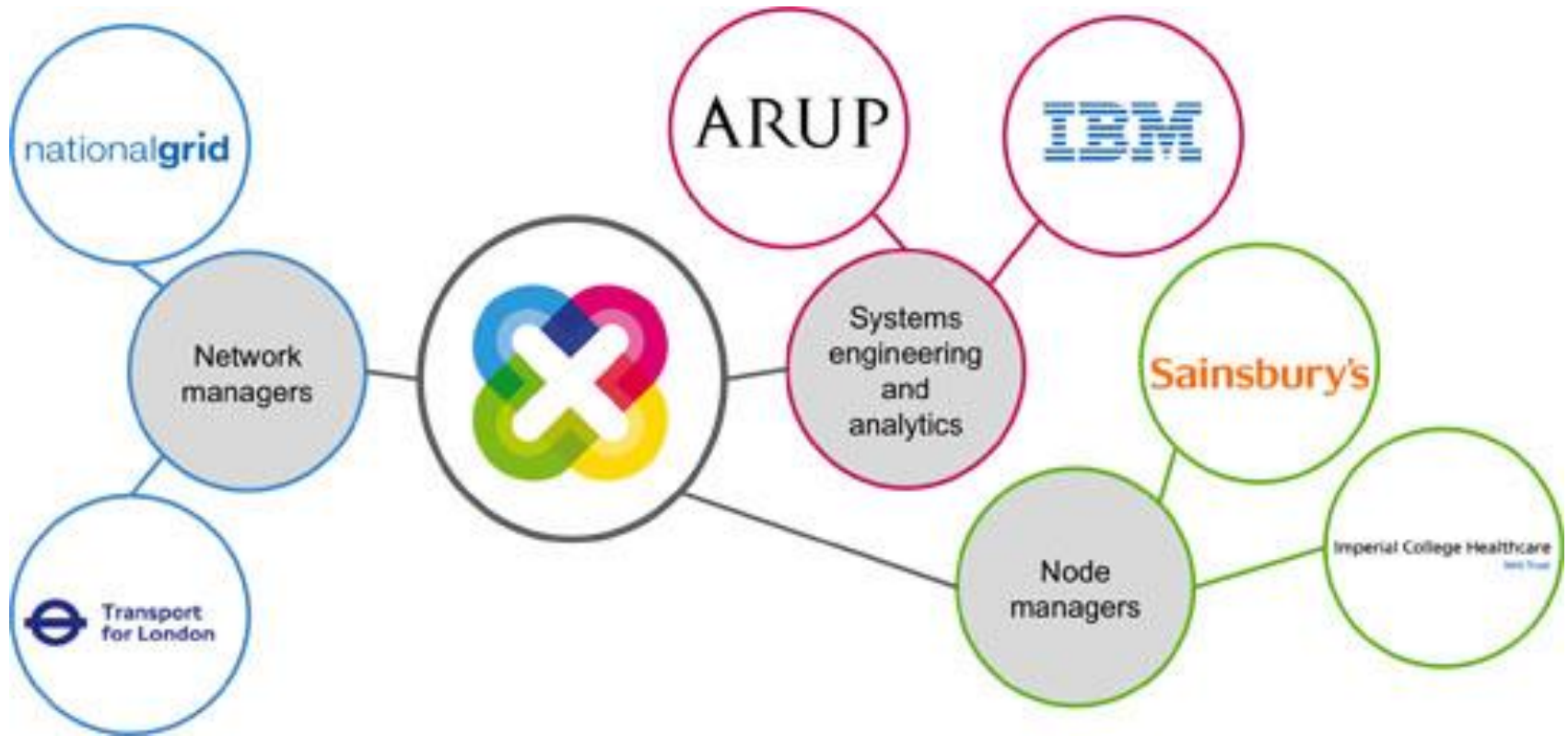
- Sector-specific data aggregation
- Single dimension, sector-specific services



Digital City Data to Services Routes

- Multi-sector Integrative Layer
- Multi-dimension, cross-sector services



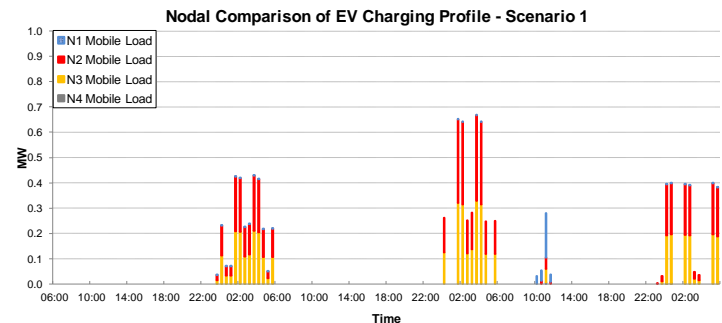
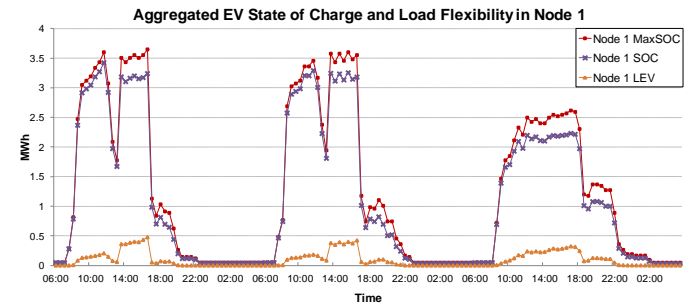
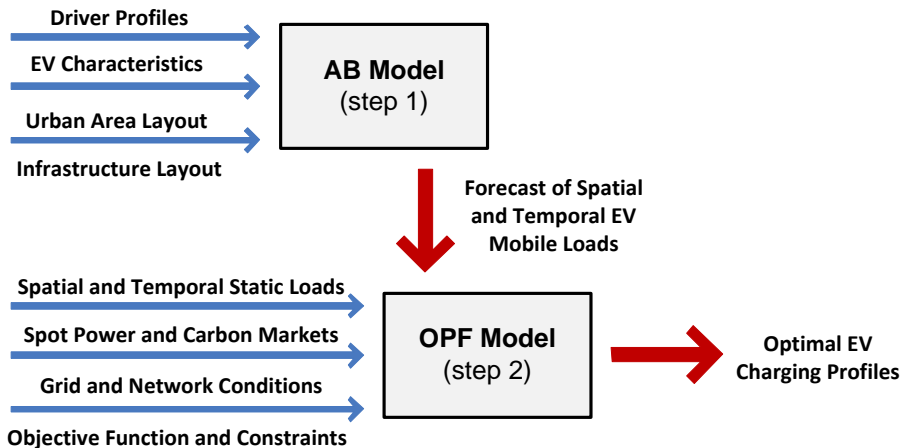


Electric Vehicle Case

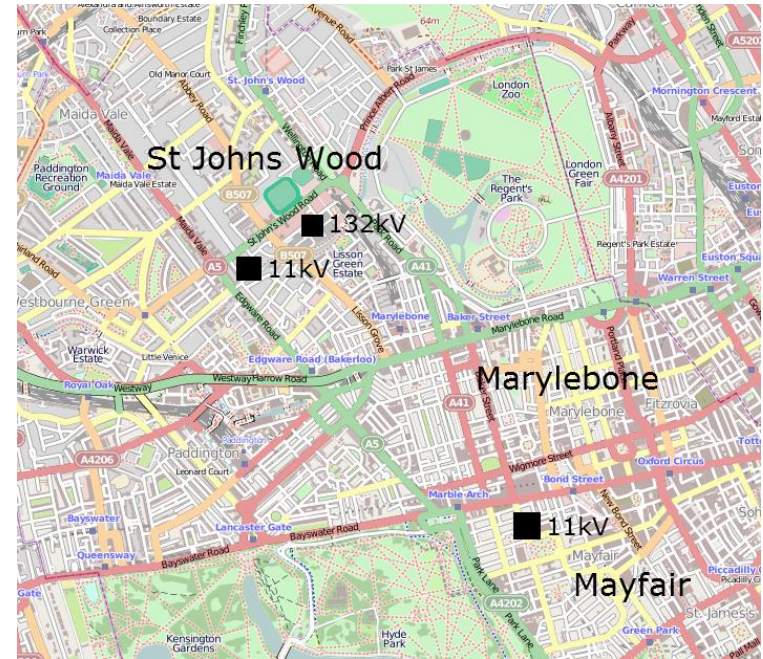
- Determining optimal charging of electric vehicles (EVs) is key in developing an efficient and robust smart-grid
- Need to **understand** vehicle movements and **predict** demands to **analyse** impact on grid and **optimise** charging profiles
- Link energy and transport infrastructures – a unique opportunity to test DCE concept of addressing peaks in multiple infrastructures



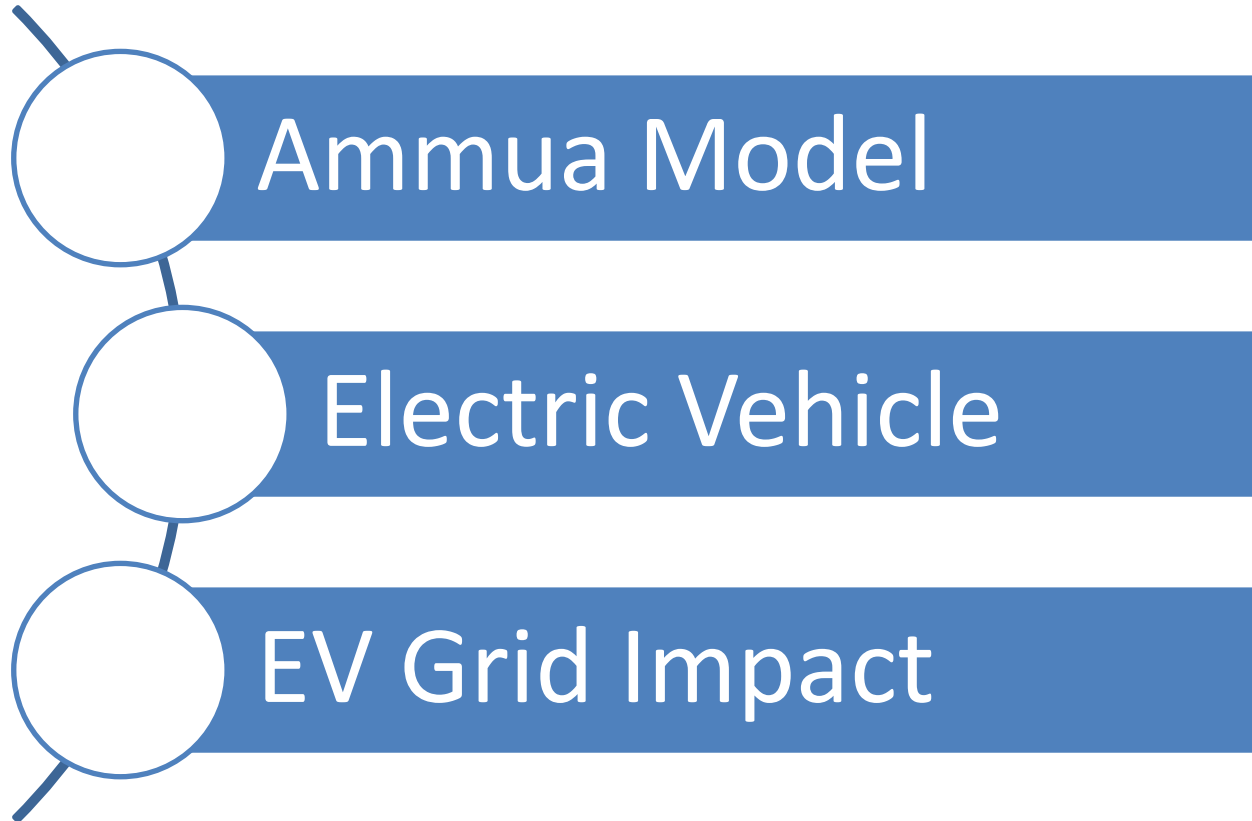
- **Phase 1:** started linking small agent-based model of EV to power flow optimiser [1]
- **Phase 2:** synthetic population of London to forecast EV movements (manually) [2]
- **Phase 3:** automate link between models [3][4]



- Two neighbourhoods in Central London with their own typical profile:
 - Residential with some retail
 - Commercial with some houses
- Predict mobile loads from EVs, fixed static loads



Metric	St. John's Wood	Marylebone/Mayfair
Number of cars	2,455	2,413
Domestic space	73%	33%
Retail space	20%	24%
Office space	7%	43%



1 – AMMUA

- **Agent based Micro simulation Model for Urban Activities (AMMUA)**
- Activity-based model simulating trips and activities in an urban environment
- Based on TASHA (Toronto, CA) and adopted and calibrated for London [5]

Inputs	Outputs
<ul style="list-style-type: none">▪ Zonal configuration of London▪ Land usage types per zone▪ Distributions of travel habits	<ul style="list-style-type: none">▪ Individual journeys from one zone to another. Includes departure and journey time.

2 – EV Model

- **Model to translate trips into EV battery status over time**
- Keeps track of people's position in the city (per zone) based on journeys from AMMUA
- For each journey the amount of energy consumed from the battery of the electric vehicle is calculated and the current state of charge (SOC) is stored

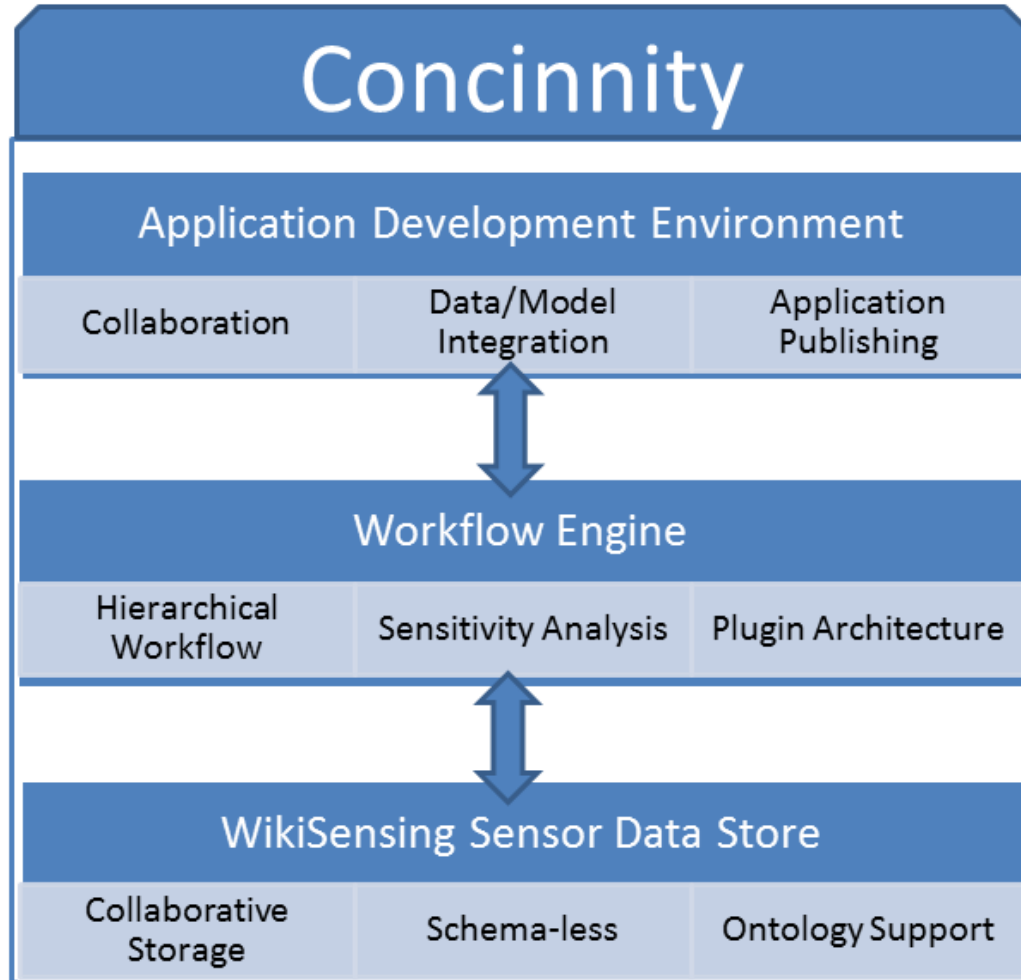
Inputs	Outputs
<ul style="list-style-type: none">▪ Trips as generated by AMMUA▪ Map of TfL zones and list of zones to study▪ EV characteristics	<ul style="list-style-type: none">▪ Snapshots at 30 minute intervals per zone with current SOC and max SOC, number of vehicles▪ Total amount of energy to charge over a 24 hour period

3 – Grid Impact Model

- **Time-coordinated power flow optimiser minimising energy or emission costs incurred from charging EVs.**

Inputs	Outputs
<ul style="list-style-type: none">▪ EV battery status over time and space▪ Static energy loads, retail and office floor space, number of cars owned*▪ Grid conditions, including carbon and electricity spot prices▪ Objective function and constraints	<ul style="list-style-type: none">▪ Optimised load profiles per substation▪ Costs (£)▪ Emissions CO₂

DCE platform



See [6]

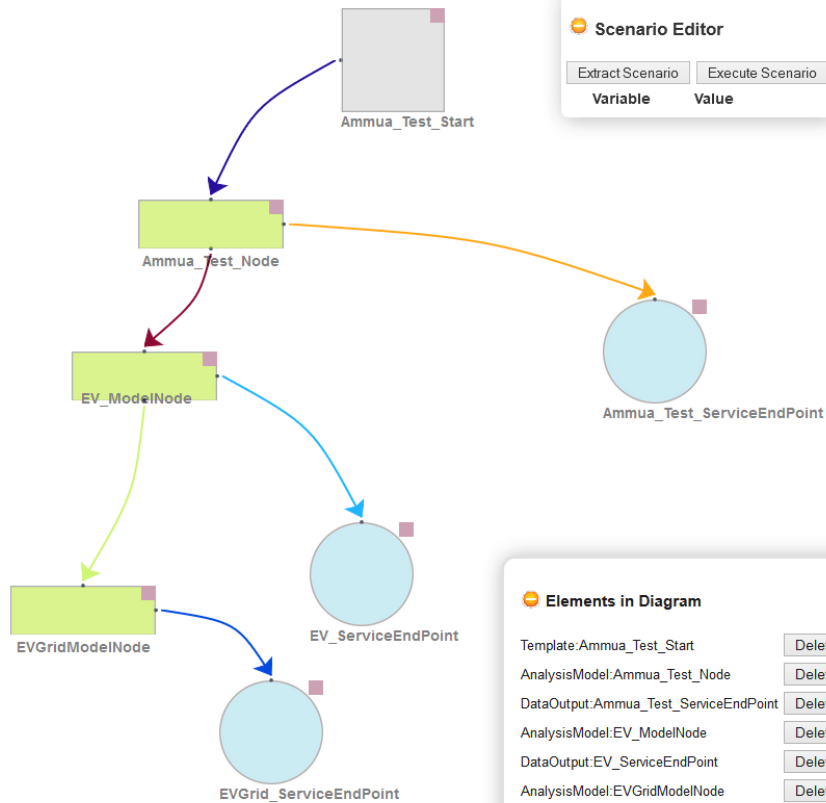
Available Elements

- Element Templates
 - AmmuaNode
 - EVGridModelNode
 - EV_ModelNode
 - SimNode
 - GeneratorNode
 - ExcelNode
 - ExcelQueryNode
 - GISQueryNode
 - CSVNode
 - MapEndPoint
 - Stream Statistics
 - WikiSensingWriterNode
 - ServiceEndPoint
 - WikiSensingNode
- Available Elements
 - ServiceEndPoint
 - Stream Statistics
 - WikiSensing
 - Root
 - Ammua_Test_Start
 - Ammua_Test_ServiceEndPoint
 - Ammua_Test_Node
 - WikiSensingParam
 - EV_ModelNode
 - EV_ServiceEndPoint
 - EVGridModelNode
 - EVGrid_ServiceEndPoint
- Available Workflows
 - SampleWorkflow_Test
 - Ammua_Test
 - Sample_Scenario_Test
 - Ammua_EV_Test
 - Ammua_EV_Test_tree
 - EV_Study_Full

Commands

Workflow Results

Map



Scenario Editor

Extract Scenario Execute Scenario

Variable	Value

Elements in Diagram

Template:Ammua_Test_Start	Delete
AnalysisModel:Ammua_Test_Node	Delete
DataOutput:Ammua_Test_ServiceEndPoint	Delete
AnalysisModel:EV_ModelNode	Delete
DataOutput:EV_ServiceEndPoint	Delete
AnalysisModel:EVGridModelNode	Delete
DataOutput:EVGrid_ServiceEndPoint	Delete

Node Editor

Clear Tree Load Node by ID Save Save As... Delete

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Owner = [David]

IsPublic =

Node

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      "Outputs": {},
      "DataLayers": {}
    }
  
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- atmex2001
- fn_act0
- fn_000r
- fn_flow2001

Workflow Editor

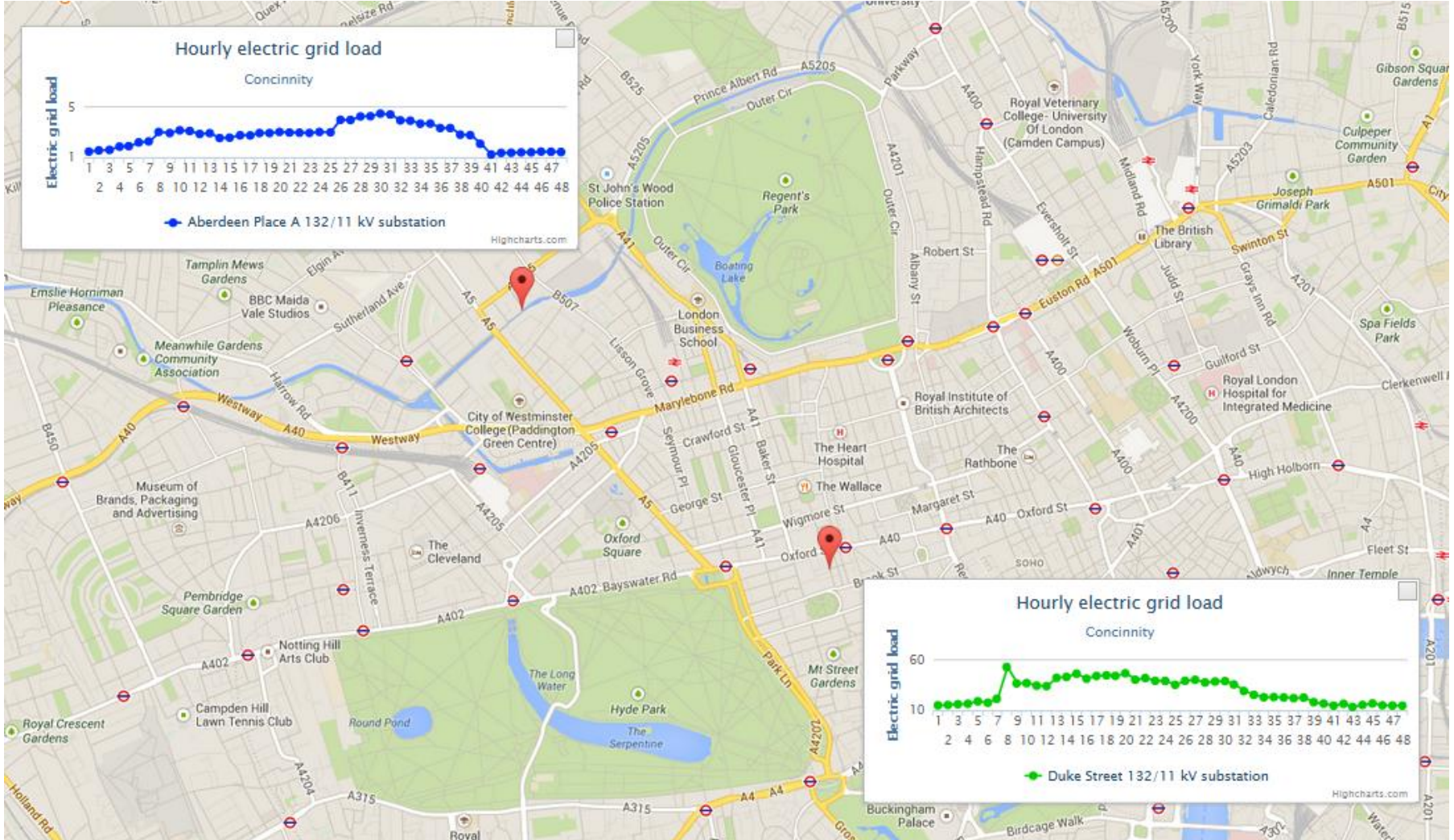
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Owner = [David]

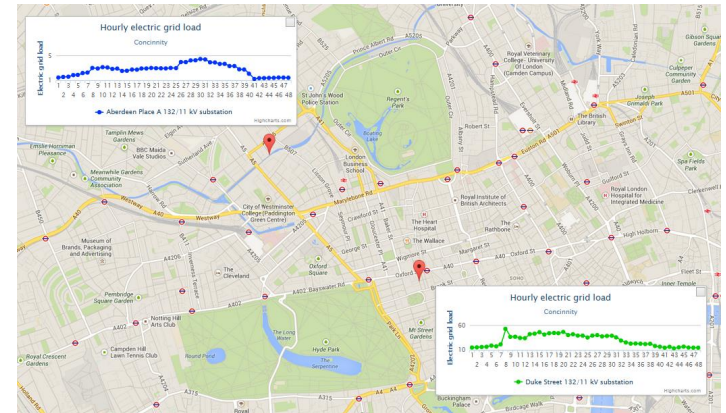
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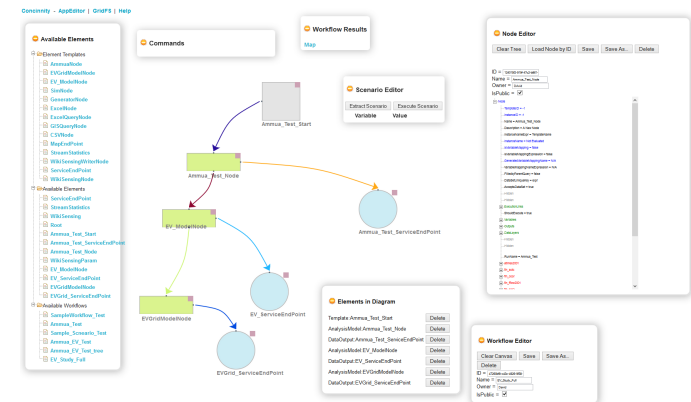
Capability	Benefits
Automation	No manual data transformation / increased accuracy by removing potential for human error / Consistency of data between models / repeatability / quick results
Collaboration	Easy access to models developed by others / collaboration / Reuse existing models in further case studies
Publication	Use of workflows, data and models by other researchers / repeatability of results / publishing of workflows with API's / Historic results storage
Scenario analysis	Sensitivity analysis of parameters Policy / demographic scenarios

Next steps

- Expanding Electric Vehicle case study shown in Concinnity Platform demo
- Idea: explore urban phenomenon at city scale looking at impact of large developments e.g. Stratford
- Two stages:
 1. Update parameters of existing models for new area
 2. Introduce additional models, incl. EV uptake, mode choice, etc



- What can we learn from this?
 - How to use the platform for larger scale studies
 - Insights city wide impact EVs
- New capabilities:
 - Expanding power flow and EV models by opening up 11kV nodes
 - Sensitivity analysis
 - Interface with data (ONS, energy prices)
 - Testing platform larger scale
 - Feedback loops (e.g. impact of energy prices on usage of Evs)
- Challenges:
 - Data on layout distribution network



- [1] Acha, S. and K.H. van Dam (2013) "Modelling Electric Vehicle Mobility in Energy Service Networks" in Modelling Distributed Energy Resources in Energy Service Networks, IET Press, ISBN: 978-1-84919-559-1
- [2] Acha, S., K.H. van Dam and N. Shah (2013) "Spatial and Temporal Electric Vehicle Demand Forecasting in Central London" in proceedings of CIRED2013, 10-13 June, Stockholm
- [3] David Birch, Orestis Tsinalis , Koen H. van Dam , Chun-Hsiang Lee, Dilshan Silva, Chao Wu, Moustafa Ghanem, Yike Guo (2013) Concinnity: A Digital City Exchange Platform, proceedings of DE2013: Open Digital, 4-6 November, Salford, UK
- [4] Koen H. van Dam, Salvador Acha, Aruna Sivakumar, John Polak and Nilay Shah (2012) Smart cities through data, models and services -- a model exchange platform, DE2012: Digital Futures, October 23rd - 25th 2012, Aberdeen, UK
- [5] Sivakumar, A., Vine, S. L. and Polak, J.W. (2010) An activity-based travel demand model for London. In Proceedings of the European Transport Conference, Glasgow, UK, October 2010.
- [6] Chun-Hsiang Lee, David Birch, Chao Wu, Dilshan Silva, Orestis Tsinalis, Yang Li, Shulin Yan, Moustafa Ghanem, and Yike Guo (2013) Building a Generic Platform for Big Sensor Data Applications. 2013 IEEE International Conference on Big Data



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