



Sizztech
ACHIEVING INNOVATION

2015 ABNMS Conference

Application of Bayesian Networks in Urban Growth Models

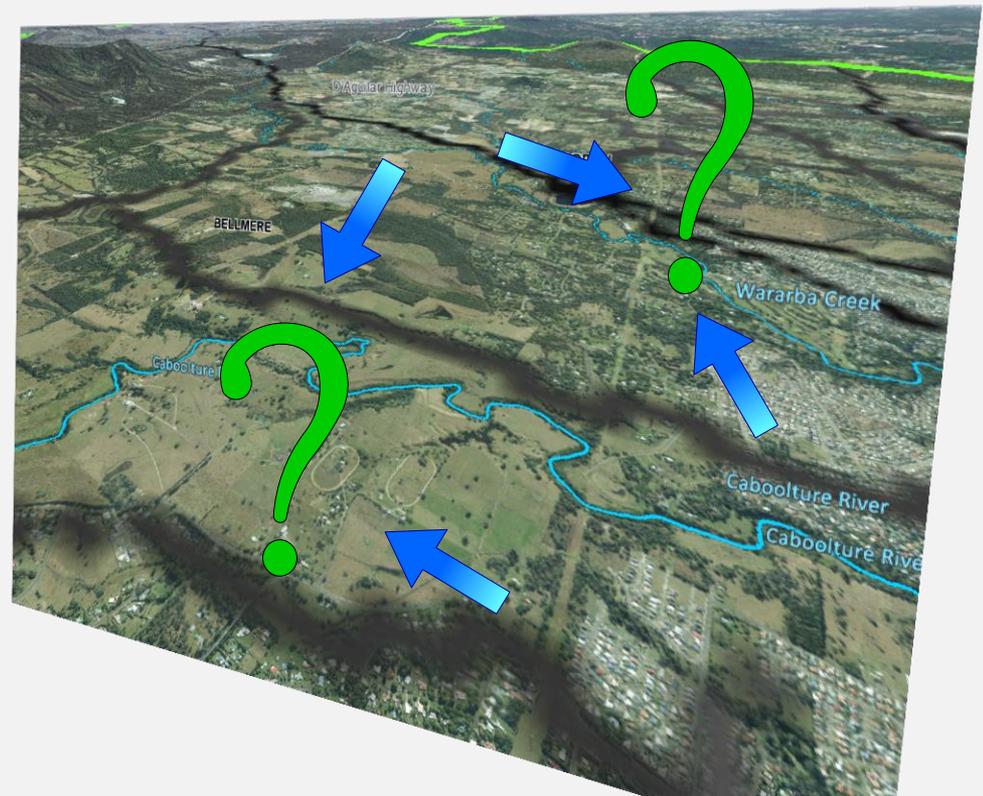
The Problem

- Majority of infrastructure projects at Local Government level
- Long term capital works programs to service future urban growth
 - Identify new infrastructure and upgrades to service urban growth for next 20 years
 - Required for all “trunk” infrastructure networks (water, sewerage, transport, stormwater, open space)
 - Associated financial models required to fund and deliver infrastructure
- Requires reliable urban growth models
- Producing urban growth models is difficult
 - Manually intensive typically taking 5-8 man months with external consultant involvement
 - No consist methodology or guidelines
 - Utilises broad assumptions
 - Models that are inaccurate, difficult to modify and prone to human error

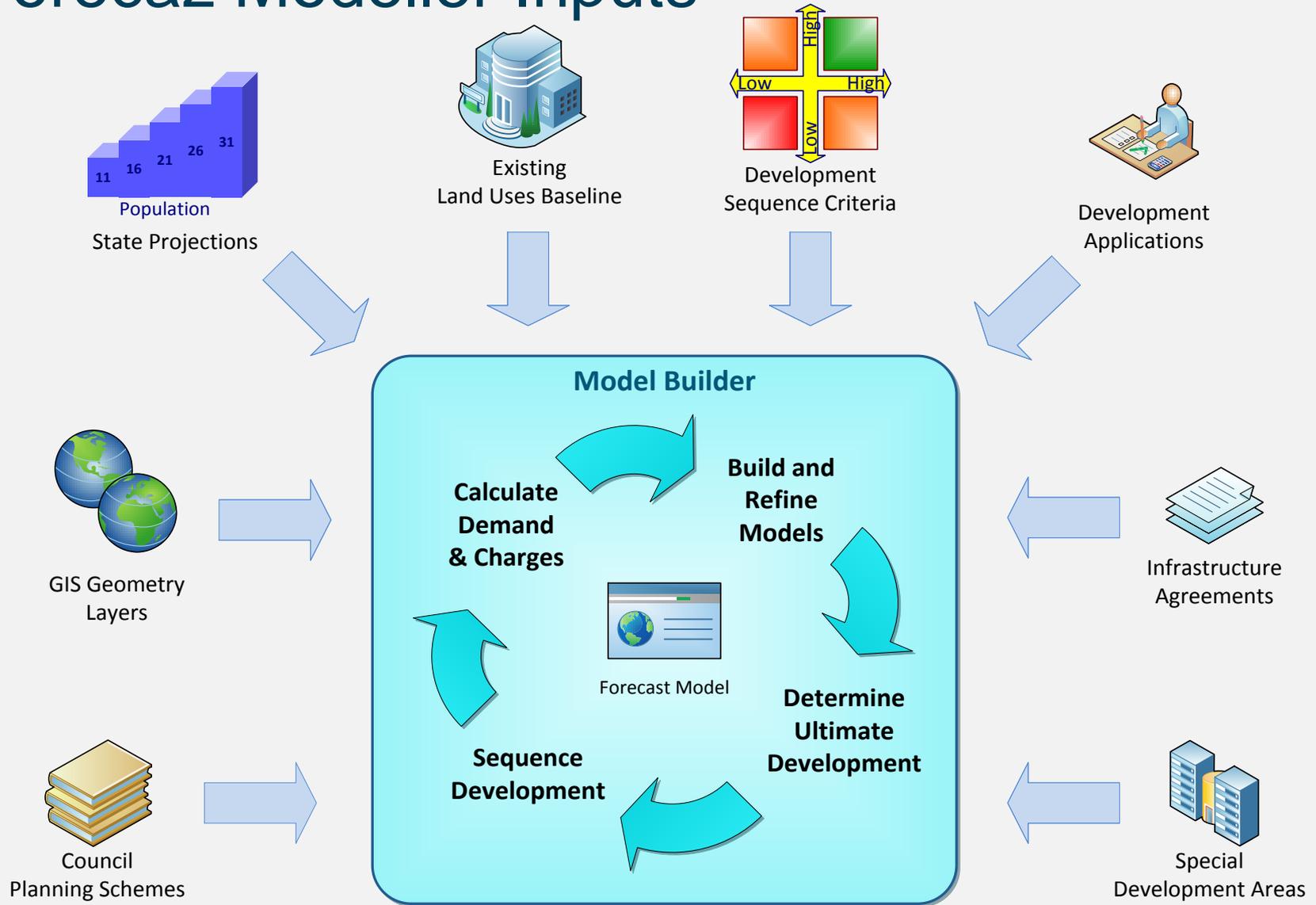


Forecaz Modeller

- Forecaz Modeller developed to address this problem
 - Geospatial modelling tool that automates the generation of residential and non-residential development growth for a region
 - Provides demand growth assumptions for infrastructure networks and calculates infrastructure charges revenue generated by the development.
 - Geospatially aggregates and views forecast model data
 - Predicts
 - What type of growth
 - Location of growth
 - How much growth
 - When will it occur
 - When is important
 - \$15M project deferred 1yr can result in \$1M of savings



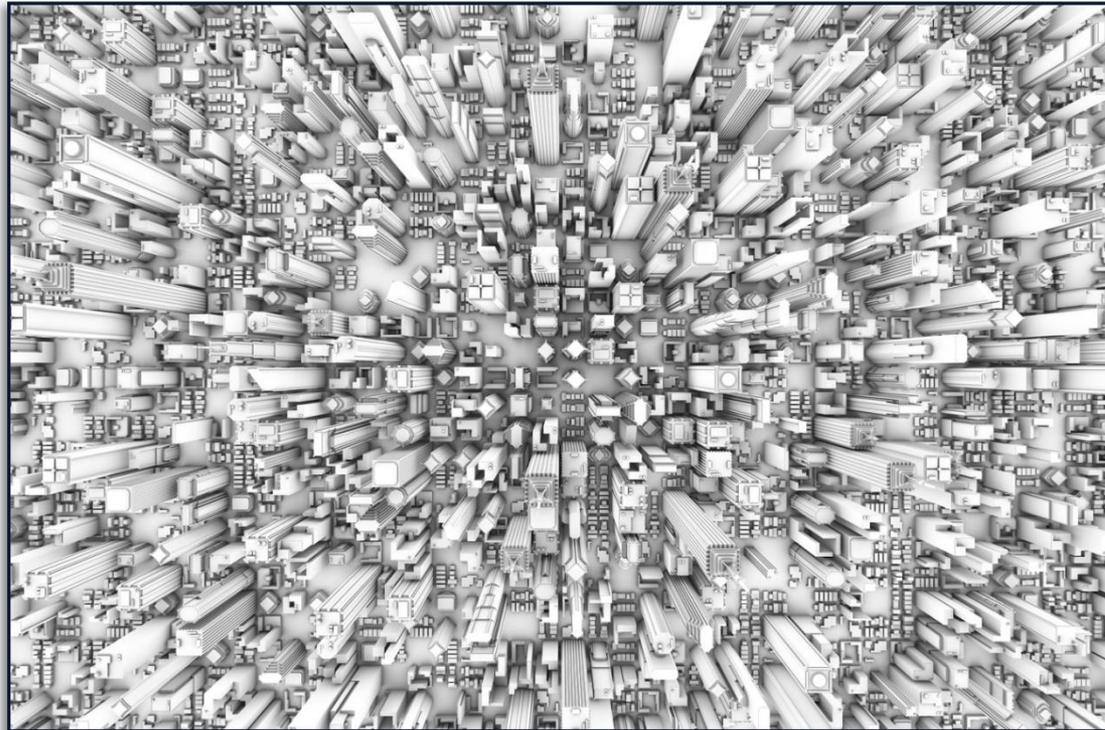
Forecaz Modeller Inputs



Forecaz Modeller

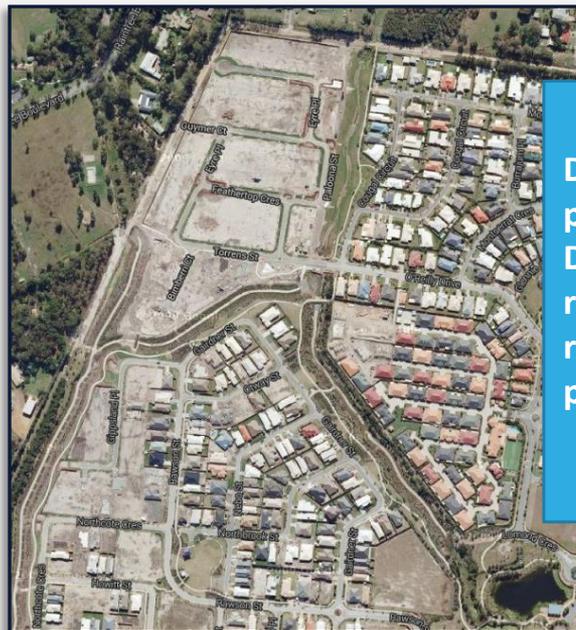
Determine Ultimate and Best Use

- Use planning zone density provisions and constraints to determine the ultimate development that can be achieved on land parcels.
- Compare existing baseline use with ultimate use to determine highest and best use.
- Determines if land parcel can be developed and the quantum of development.



Forecaz Modeller Sequence Development

- Utilise population and m2GFA growth projections as “control” targets for residential and non-residential development growth at specified projection years.
- Modeller employs a **Development Desirability Index (DDI)** indicator of the desirability for a developer to develop a land parcel.
- Utilise DDI to allocate and sequence developable land parcels to ultimate full development within projection years until control targets are reached.



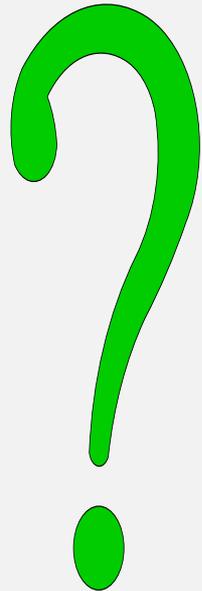
Develop land parcels with highest DDI until res & non-res control targets reached for a projection year



Modeller's BN

Why Bayesian Networks

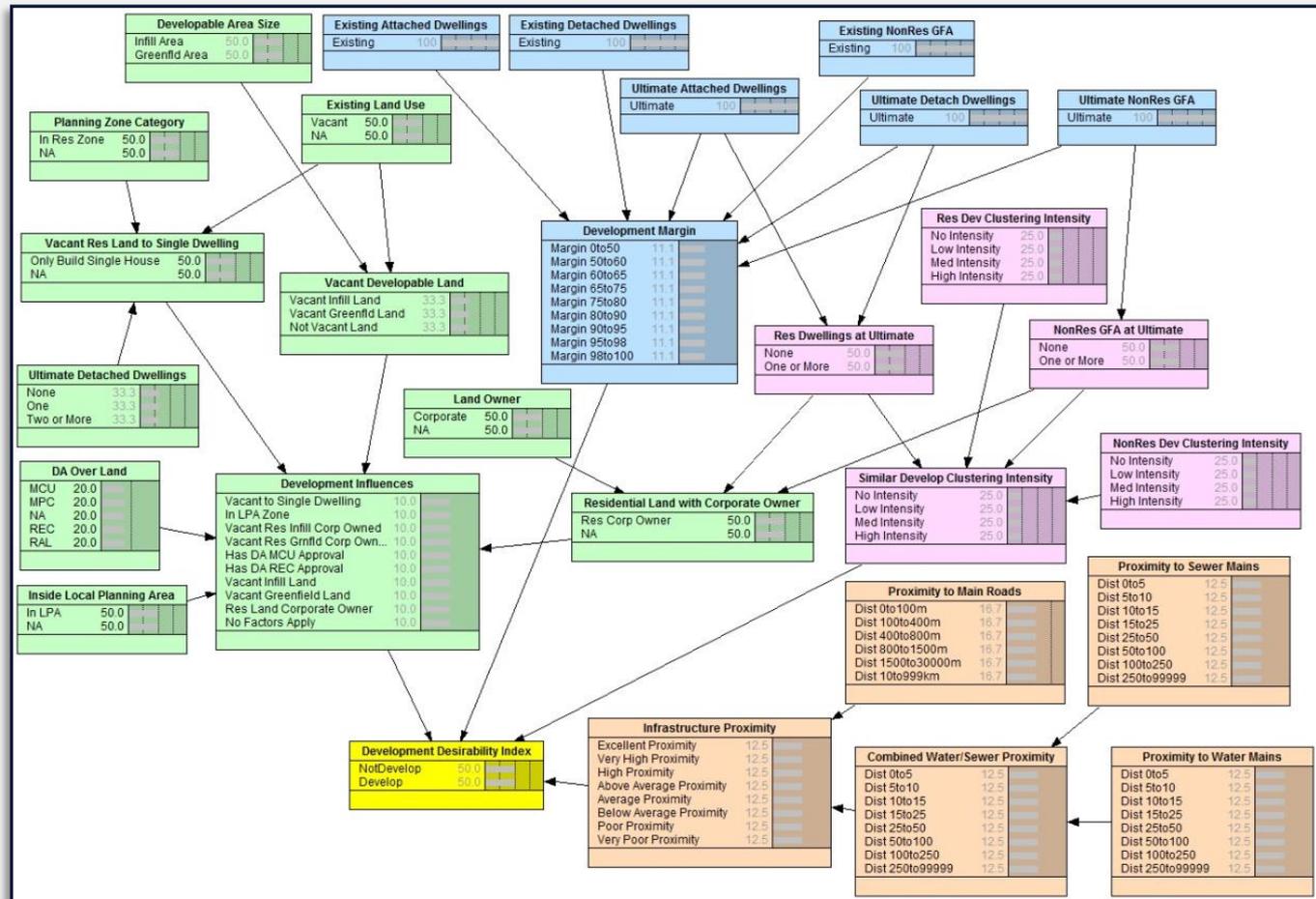
- Utilises a BN Model to describe a predictive development algorithm that determines a land parcel's **Development Desirability Index**
- Why use Bayesian Networks?
 - Graphical models that have a clear schema and conditional probabilities that are easily understandable by non-experts
 - BNs incorporate the uncertainty of expert knowledge
 - Primary users are Urban Planners, safe to assume no Post Grad Mathematical Degree to maintain complex predictive models
 - The methodology used to determine growth assumptions needs to be public available and understood by a lay-person
 - The BN tools are mature, have a strong community, well supported and documented and have a comprehensive API



Modeller's BN

Bayesian Network Submodules

- Development Margin
- Development Influences
- Infrastructure Proximity
- Similar Development Clustering



Modeller's BN Assumptions

- Only land parcels that can be developed are presented to the BN Model
 - No BN consideration for: flooding, reserves, fully developed land, etc.
- No accounting for proximity to Community Facilities (i.e. shops, parks, schools)
 - Assumed these factors taken into consideration when local governments construct their planning schemes and master plans
- No environment considerations are provided for by the BN Model
 - Development constraints are applied through a “constraints” GIS layer by the Modeller
 - Only parcels with developable area are presented to the BN Model

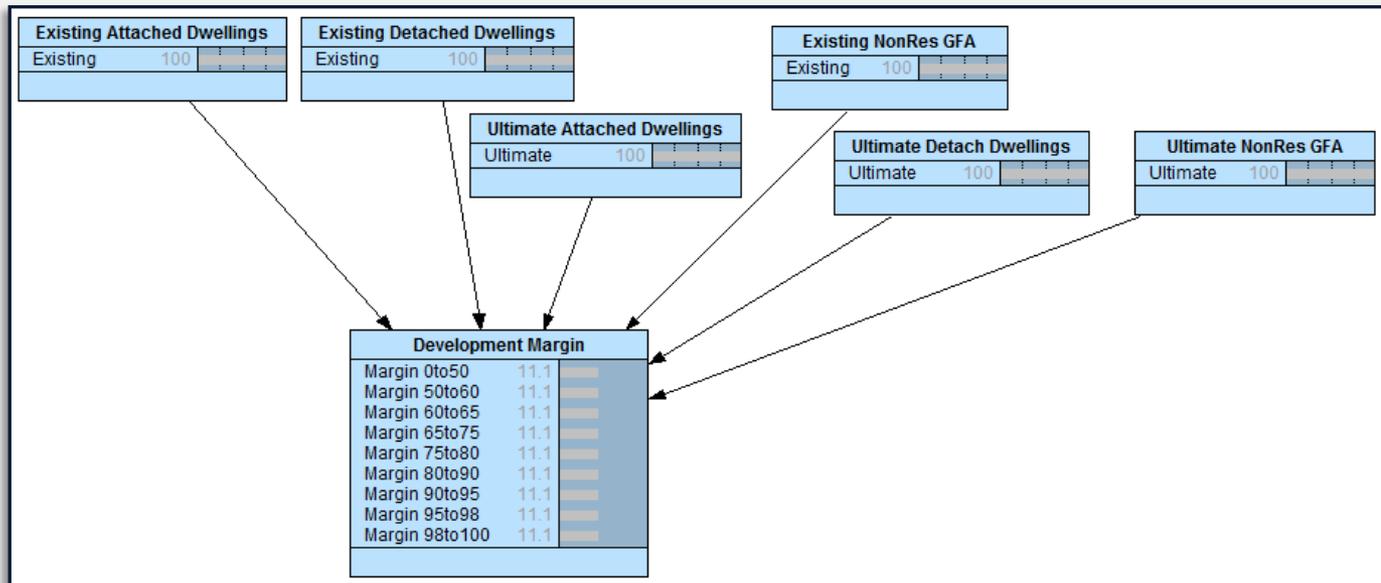
Modeller's BN

Development Margin Submodule

- Calculates the development margin being:

$$\frac{\text{Ultimate Development} - \text{Existing Development}}{\text{Ultimate Development}}$$

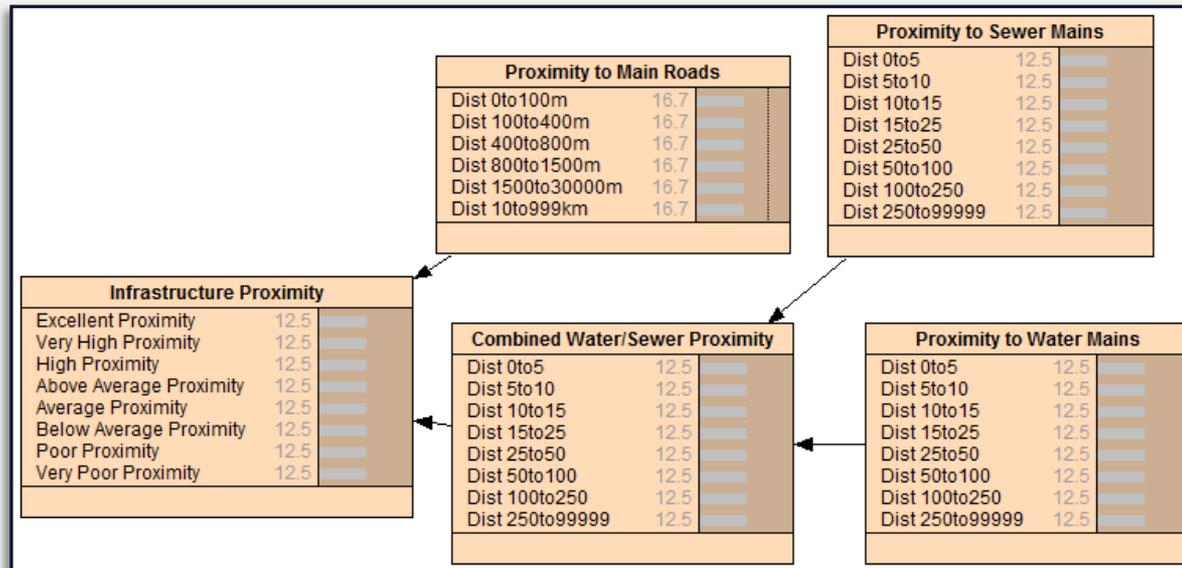
- Node discretisation has been structure to transpose development margin to Development Yield. “Margin 0to50” represents poor yield of less than 200%
- Development Yield is a primary development factor. High yield -> high probability of development.



Modeller's BN

Infrastructure Proximity Submodule

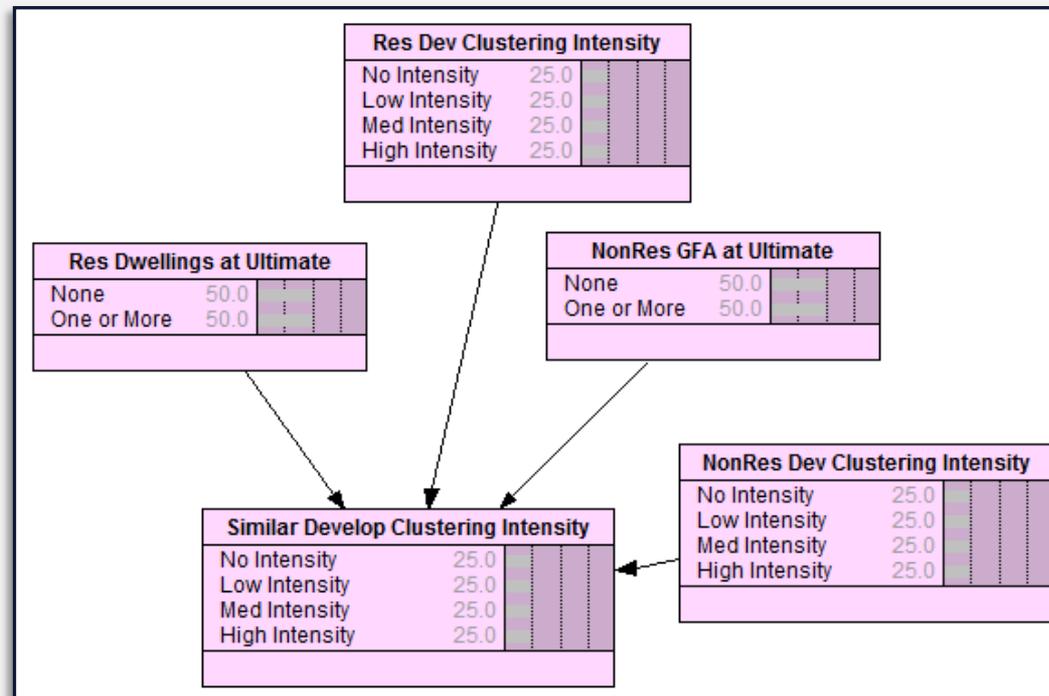
- Combines proximity to Water, Sewerage and Main Roads infrastructure into a single distance state representing proximity to network infrastructure.
- The node is weighted towards the proximity to water mains. A water connection is required, before a Land Parcel can have a sewerage connection.
- Infrastructure Proximity is a primary development factor. Developers face significant costs to expand trunk infrastructure to service their development.



Modeller's BN

Similar Development Clustering Submodule

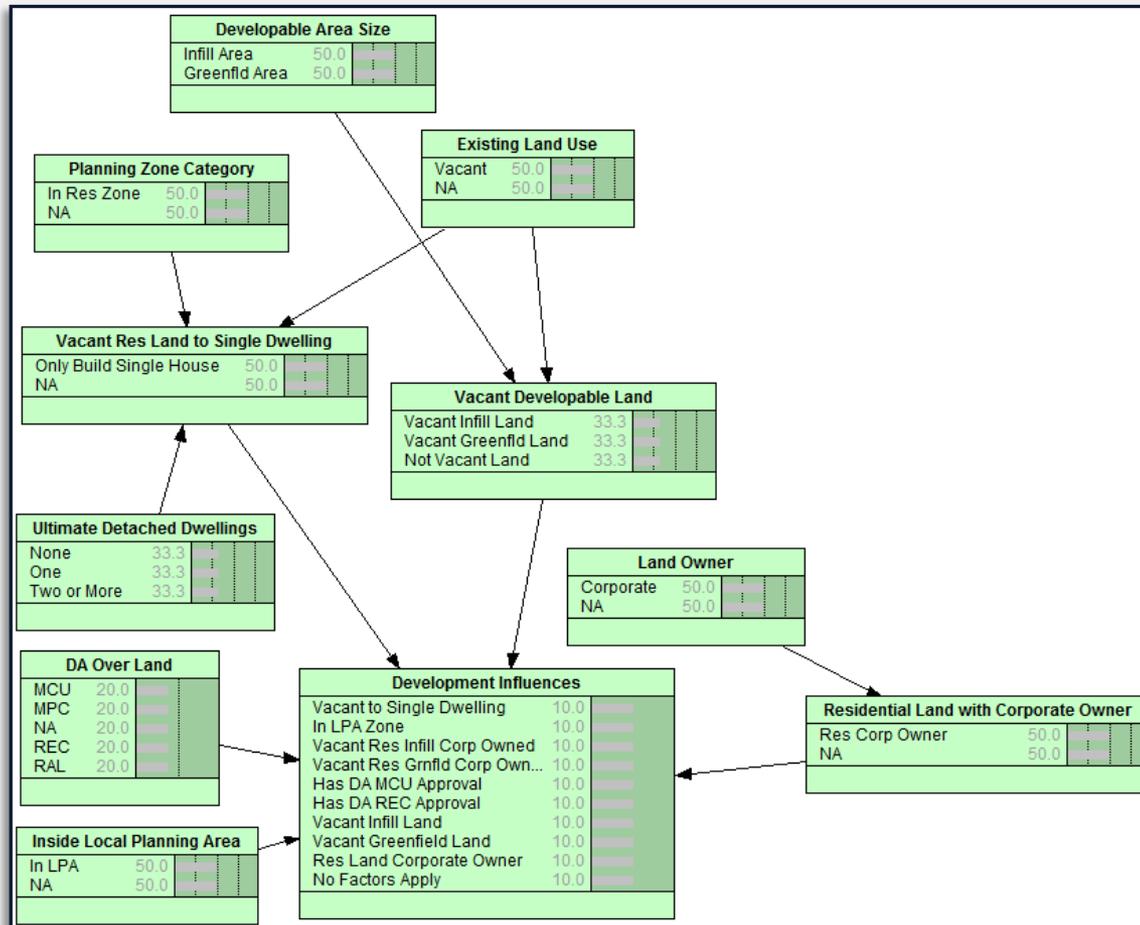
- Represents a similar development clustering/precedence within 100m from boundary of a land parcel.
- This node combines the states of residential and non-residential similar development clustering
- Only the clustering types (res / non-res) that matches the land parcel's ultimate development are considered.



Modeller's BN

Development Influences Submodule

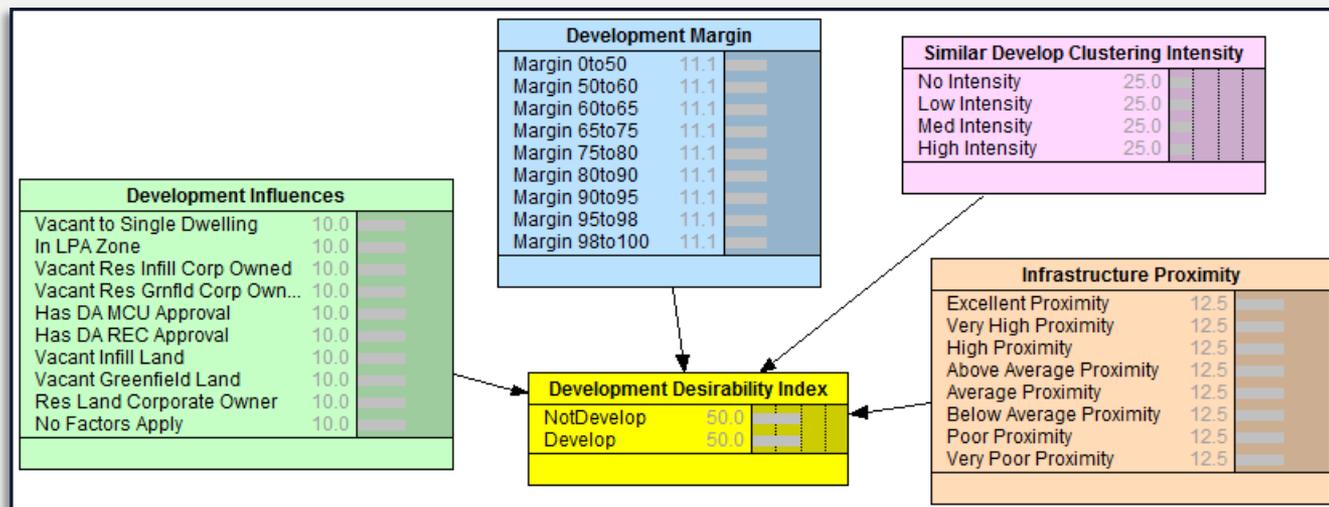
- This node combines the factors that influence development from its various input nodes and weights them as discrete states for input into the Development Desirability Index node



Modeller's BN

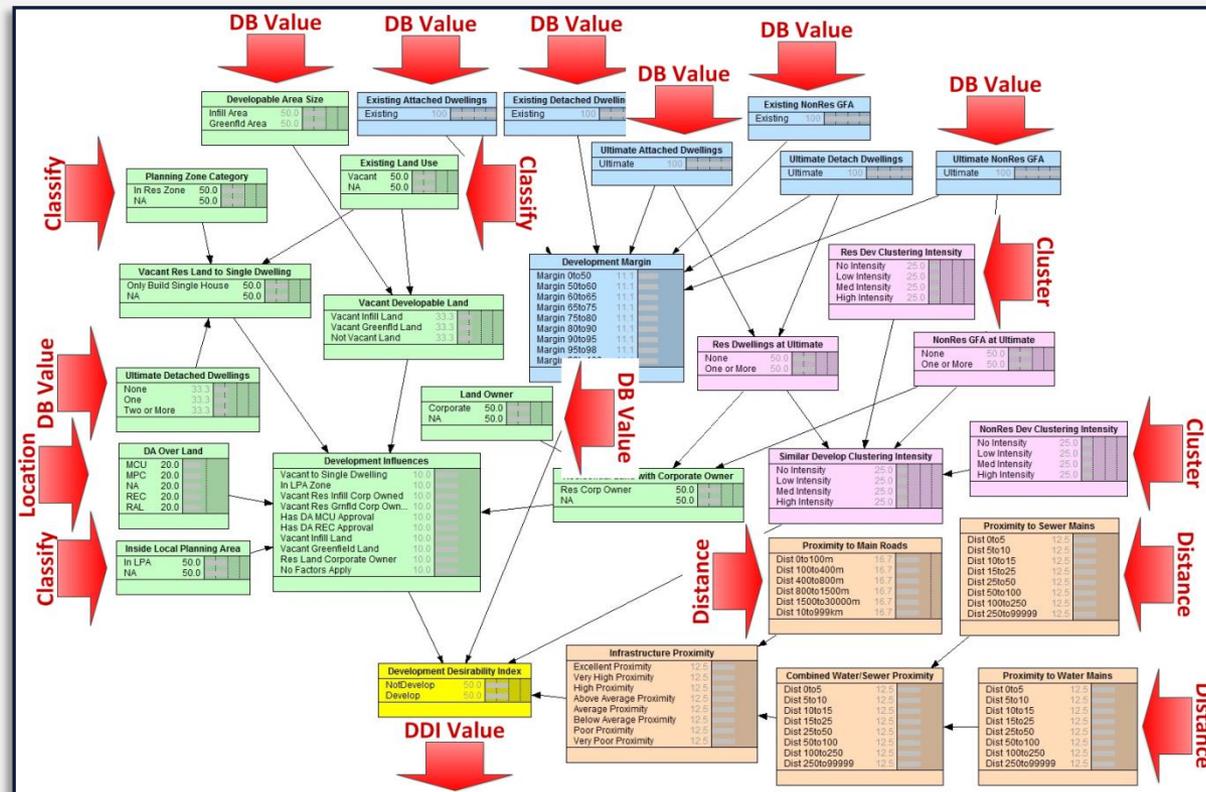
Development Desirability Index Node

- This Node contains the Development Desirability Index (DDI) for the Land Parcel.
- Heavy weighting is given to **Development Margin** and **Infrastructure Proximity**
- The highest DDI occurs when **Development Influences** has a state of “*Vacant to Single Dwelling*”. Represents vacant land created from a recent sub-division.
- Forecaz Modeller extracts the DDI from the BN Model after: populating all nodes; computing any equations within model; compiling the model to obtain the model's beliefs which produces the DDI for a developable land parcel.



BN Integration Overview

- The BN nodes are populated by extracting and filtering data from the database or performing spatial operations.
- The Modeller provides computation methods that can be configured to extract and correctly format data for nodes.



BN Integration

Computation Methods

- **Development Desirability Index Method**
 - Identifies the node and state that contains the DDI Value
- **Database Value Method**
 - Extracts and filters values from the database
- **Distance Method**
 - Minimum distance between parcel and GIS layer features
- **Located Method**
 - Parcel located within a GIS feature returns an attribute from feature
- **Classify Method**
 - Classify various values of an attribute in an intersected GIS feature into a single classification value
- **Cluster Method**
 - Count of parcels meets Res/Non-Res thresholds within specified distance from parcel

Next Steps

- Used the modeller to build urban growth models for:
 - Noosa Shire Council
 - Sunshine Coast Council
- Currently developing Moreton Bay Regional Council model
- Take 3 data sets and apply learning from data to further validate the BN
- Use learning to expand the nodes (i.e. distance to bus stops/train stations) and evaluate sensitivity
- Investigate the use of Decision Networks



Thank You