#### A MODERN APPROACH TO LAND VALUATION: AN APPLICATION OF ARTIFICIAL NEURAL NETWORKS

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## Motivation

- Artificial intelligence (AI) techniques have gained increasing attention over the past few years.
- Artificial neural networks (ANN) have been used for various purposes, including prediction, pattern recognition, process control and data analysis.
- The advantage of the ANN approach is that it can examine complex nonlinear relationships without many of the parametric restrictions and a priori assumptions about the data and functional form used in conventional regression models.
  - A growing literature on the use of ANN in real estate research (Peterson and Flanagan 2009; Lin and Mohan 2011; Zurada, Levitan and Guan 2011; McCluskey et al. 2013; Yacim and Boshoff 2018)

# Background – ANN

- Artificial neural networks provide a learning system consisting of a large number of neurons.
  - Each neuron can process signals received and signal neurons connected to it.
  - With neurons organized in interconnected layers, ANNs can emulate almost any function and solve various problems, assuming sufficient training data and computing power.
- The power of a neural network is determined by the activation functions, the learning rule, and its architecture.

## Background – ANN Structure

Figure 1: An Example of ANN Architecture in Real Estate Studies



# Background – ANN Training Process and Algorithms

- Available data divided into three sets: one for training, one for validation and the rest for testing.
  - Using the training dataset to train the neural network
  - Utilizing the validation dataset to evaluate the model fit and find the best model fit
  - Reserving the test dataset for an unbiased evaluation of the selected model
- Algorithms used for training purposes: back propagation (BP), Levenberg-Marquardt (LM), and conjugate gradient (CG) algorithms
  - The BP algorithm: the most commonly used one in real estate mass appraisal based on ANN models (Yacim and Boshoff 2018)

#### Application — Land Valuation

- Application of ANN for land value estimation
  - Motivated by sales comparison approach
  - Using vacant land sales data to build ANN models, then choosing the best model to estimate land values for parcels in the improved sales dataset
- Data preparation for ANN training
  - Data cleaning and preprocessing (missing values; outliers)
  - Treatment of categorical variables (entity embeddings of categorical variables)
- Data split for ANN training and testing
  - No consensus about the data division ratio in the literature (avoid underfitting as well as overfitting the model)
  - 80% used for training purposes, and 10% for validation and testing, respectively

- A multiple-layer network
  - Multiple-layer neural networks overcoming the limitation associated with single-layer neural networks
  - Based on land value determinants in the existing literature and data availability, 45 variables used as inputs representing parcel characteristics as well as zoning regulations
- Two main hyperparameters controlling the network architecture
  - The number of layers
  - The number of nodes in each hidden layer
  - Configuring the hyperparameters through systematic experimentation with a robust test harness



- Activation functions
  - Creating complex and nonlinear mappings between input and output variables
- Popular activation functions
  - Sigmoid activation function: it has an S-shaped curve with a range between 0 and 1 (easy to understand and apply; vanishing gradient problem and slow convergence).
  - Hyperbolic tangent function (Tanh): it makes output zero centered because of its range between -1 and 1 (normally preferred over Sigmoid function; vanishing gradient problem).
  - Rectified linear unit activation function (Relu): it rectifies the vanishing gradient problem. One potential issue is the possibility of dead neurons (leaky Relu as a fix).





- In general, experimentation with ANN model configuration needs to consider efficiency as well as model capacity.
- Evaluation using the test data suggests:
  - The prediction accuracy varies across different residential and commercial market areas.
  - The prediction accuracy varies across cities.

# Application — Predictions Using New Input Information

- Predict land values for parcels in the improved sales dataset
  - Establish land values separately from property values with the constructed ANN model
  - Use parcel information from the improved sales dataset as new inputs to predict land values
- Possibility to estimate land values for parcels in the county assessment dataset
  - Similar prediction process with a few assumptions

#### Limitations

- There are general concerns about out-of-sample predictions.
  - The prediction accuracy depends on the quality of land sales data as well as the representativeness of the data.
  - Underrepresentation of certain land types would affect the predication accuracy for the types of land. (This problem can be mitigated as more land sales data become available over time.)
  - Certain land features may be important but not observable. The constructed neural networks would not be able to capture those.

#### Conclusion

- ANNs provide us with a modern approach to land value estimation. One main advantage is that these models can study complex nonlinear relationships without a priori assumptions about the data and functional form.
- They can help lessen the burden on assessors or researchers when estimating land values on a large scale.
- ANNs rely on a large amount of training data to appropriately solve various problems. Inadequate information would result in poor predictions.
- Potential overfitting problems affect the generalizability of ANN models. Building a model that generalizes well to new data is always challenging.