



# Evidence-Based Transportation Demand Analysis

## *Evidence, Model & Parameter Estimation*

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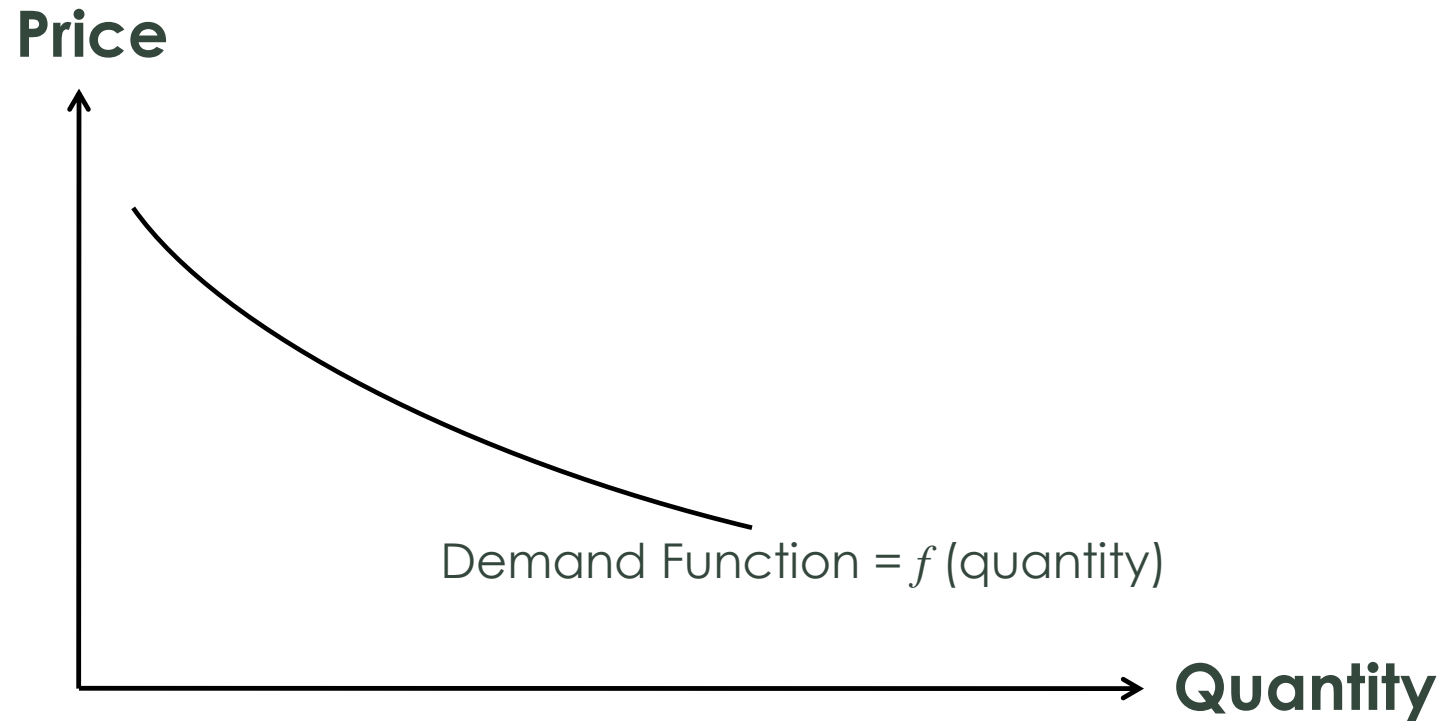
# Outline

- Economic concept of transportation
- Data/Evidence and errors
- Sample versus population
- Econometric model

# Economics of Transportation Demand

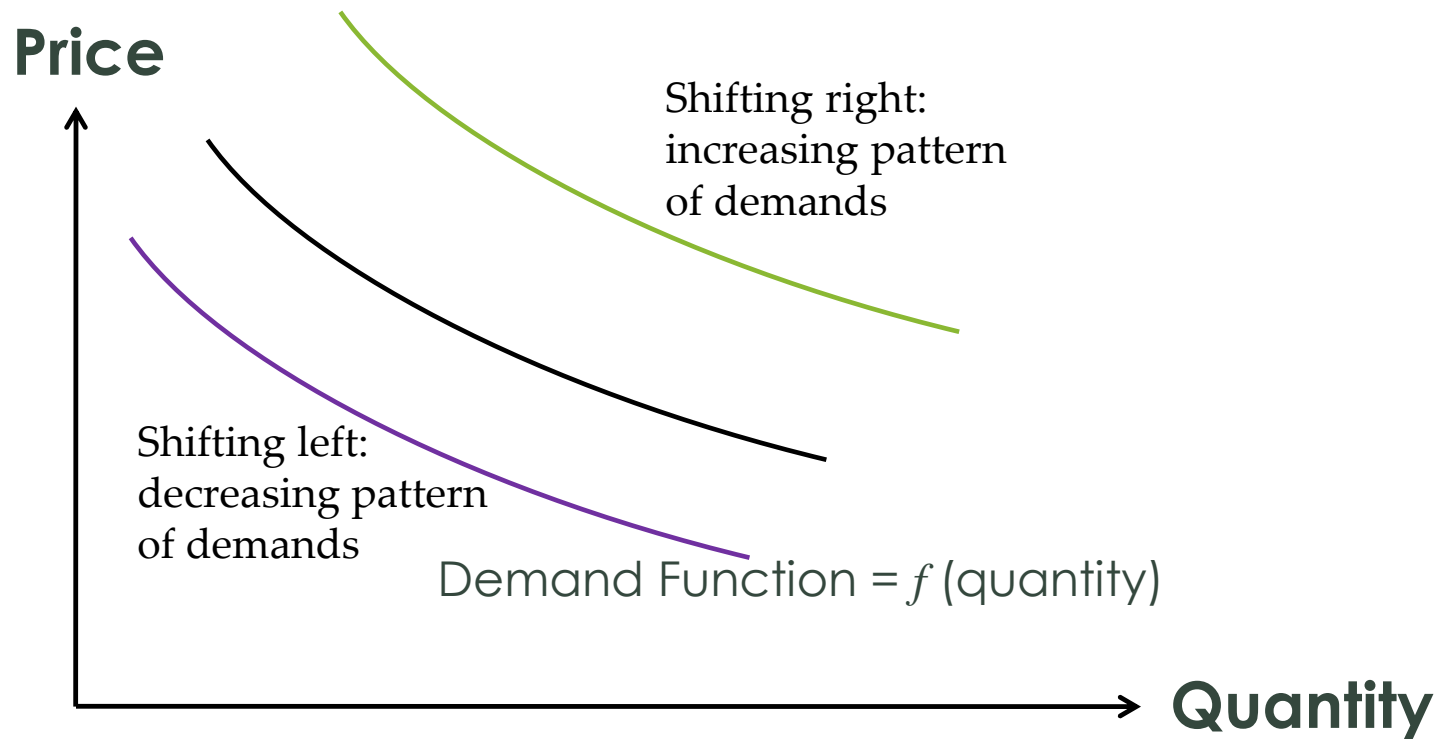
- Key concepts:
  - Demands for transportation: People generates such demand that are derived from the need to engage in activities at different locations
  - Transportation system enables the movements and thereby supply the means of transportation at the cost of travel time, fare/fuel cost, charges (toll, parking, etc.) and other externalities
  - Demand-Supply interactions:
    - Price/cost equilibrium
    - Externalities: congestions/delays, (in)convenience, (un)reliability, collisions/accidents, emission, pollutions, etc.
    - Consumer surplus/Social welfare

# Demand Curves



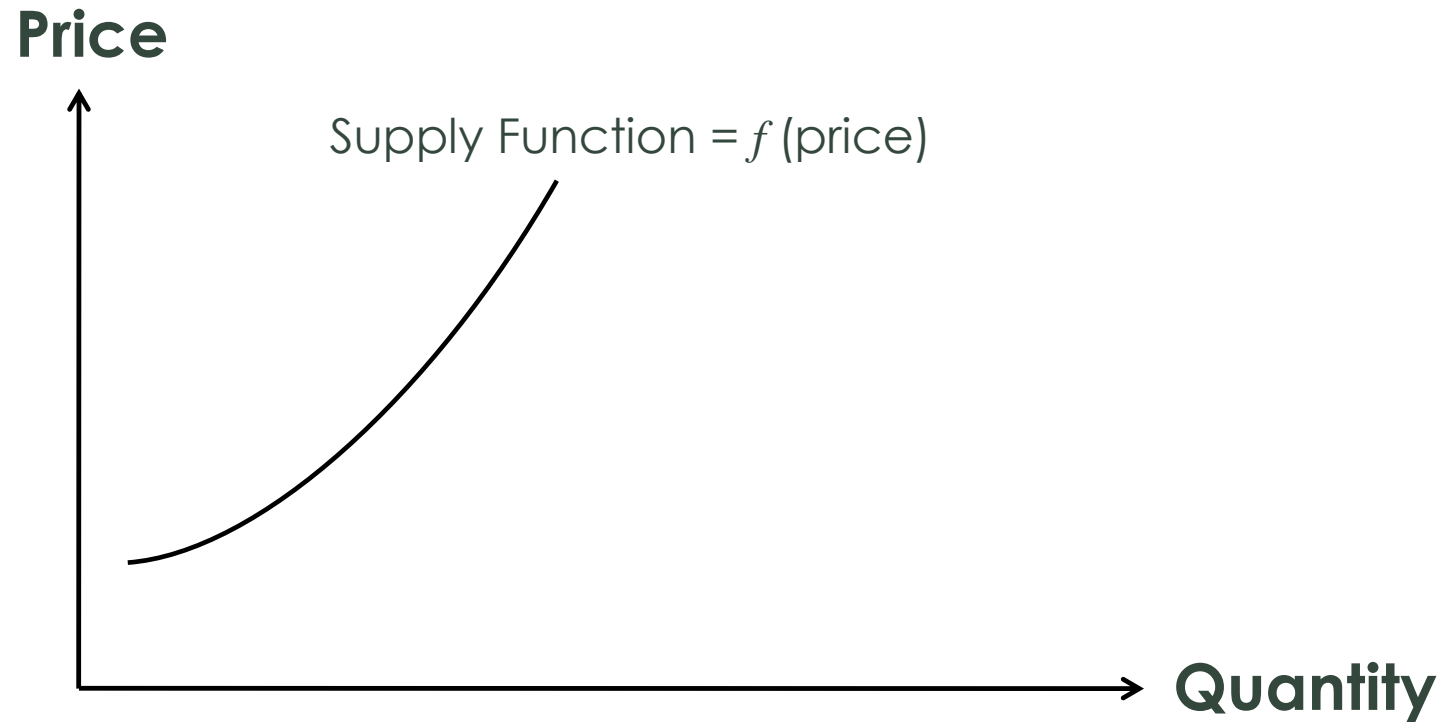
- Price/Cost
  - Travel time, cost, safety, comfort, convenience, etc.
  - Externalities
- Quantity of Demand
  - Total number of trips per day
  - Trips by different purposes
  - Trips at different destinations
- Demand Curve
  - Demand function
  - Cost/Price sensitivity of total demand

# Changing Demand Patterns



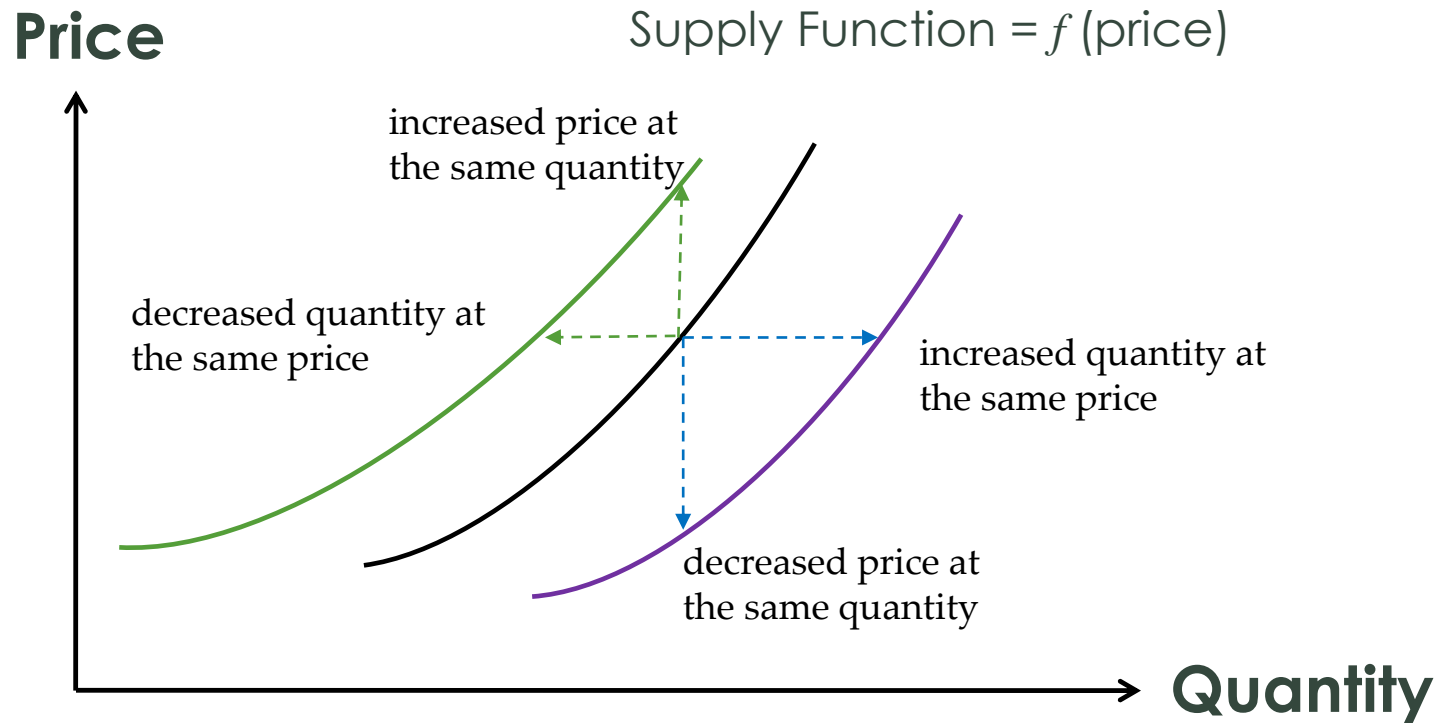
- Price sensitivity: Changes in quantity of travel due to changes in other variables than perceived price/cost of travel:
  - Income
  - Land use and urban form
  - Lifestyle, technology
  - Population and regional economy

# Supply Curve



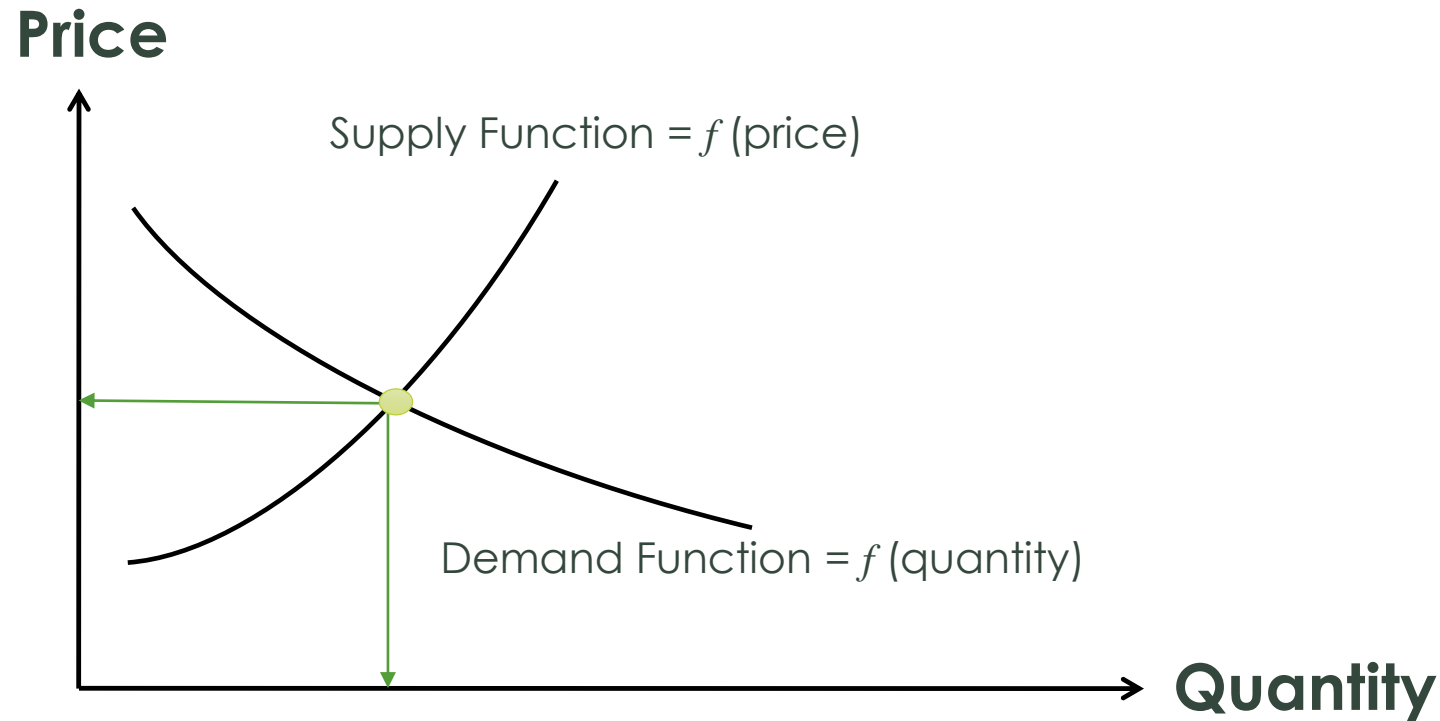
- Price
  - Average travel time, congestion delay, externalities, etc.
- Quantity of demands
  - Flow, density, speed
- Supply curve
  - Performance function

# Changing Supply



- Increase in system Capacity: shifting to the right
  - Adding new element
  - Efficient operation and maintenance
- Decrease in system Capacity: shifting to the left
  - Reducing elements
  - Deterioration

# Demand-Supply Equilibrium



- Equilibrium: the observed system
  - Observed link flows, speed, average travel time, average travel cost etc.
  - Observed total number of trips by zones, households, persons etc.
  - Observed trip distribution
  - Observed modal splits



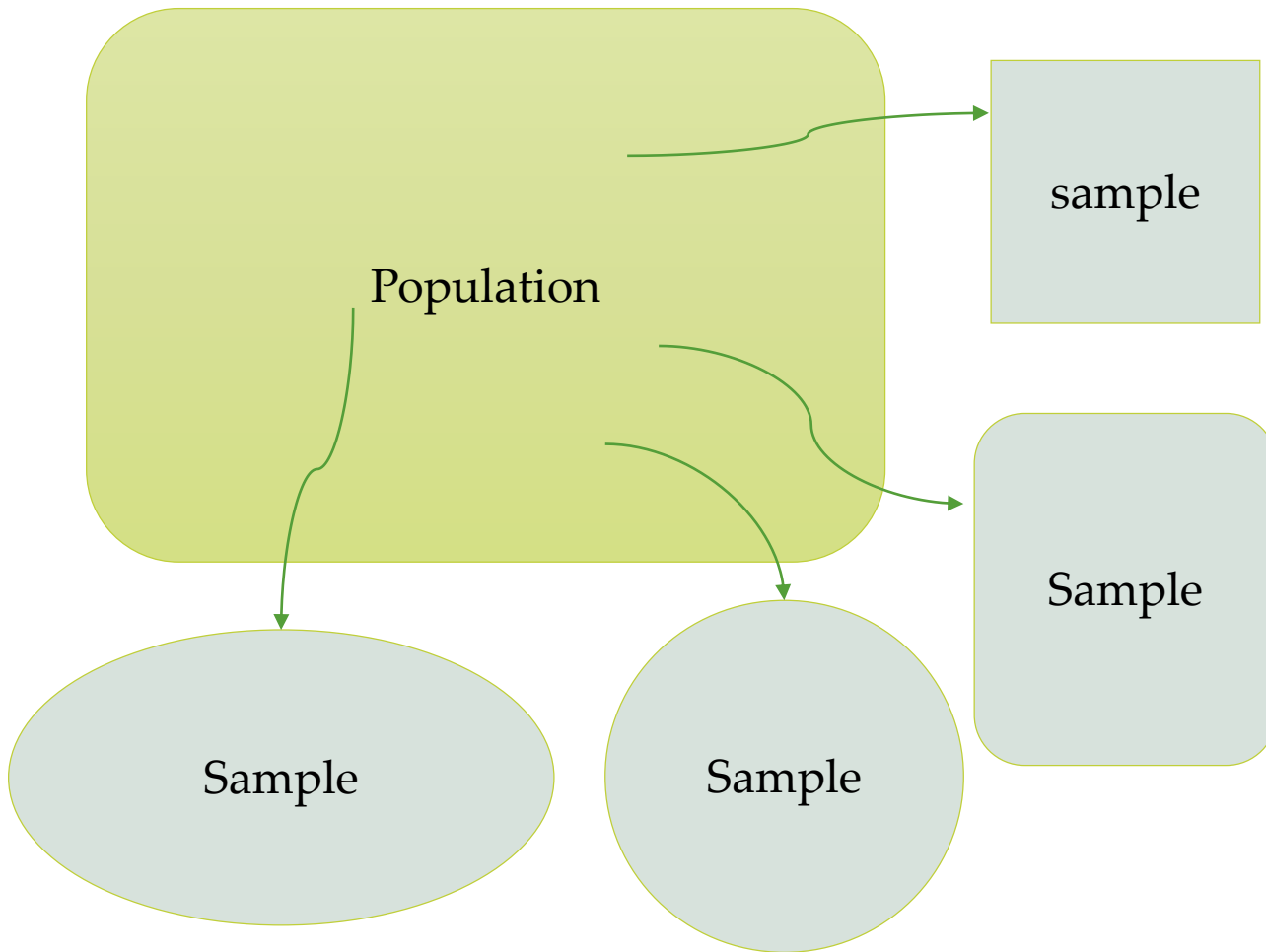
# Evidence of Transportation Demand

- Evidence = Observed reality or Stated opinion/choice/preferences
- Transportation system observation (supply) has to have boundaries:
  - Observation time frame: daily observation, weekly observation, etc.
  - Observation extent: system wide observation, observation of key links, routes, corridors, etc.
- Transportation demand observation has to have boundary:
  - Observation of whole population of travellers (missing people not travelling) on a particular time-of-day, on a particular day
  - Observation of a sample of population (sample of households including those not generating any trips) on a particular day, week, or longer

# Evidence of Transportation Demand

- Constraints in evidence generation (observation)
  - **Temporal constraint:**
    - Impossible to observe for indefinite time period
    - So, even a comprehensive observation within a finite timeframe is a sample of infinite temporal dynamics
  - **Spatial constraints:** Urban space and network need to be categorized for meaningful data presentation:
    - Zones, planning district etc. are sample of urban space
    - Highways, arterials, corridors etc. are sample of network
  - **Population/Socio-economic constraints:**
    - Even a census is a sample of continuous socio-economics and population dynamics
    - Majority of time, we rely on sample of households or individuals

# Evidences are Samples of Reality

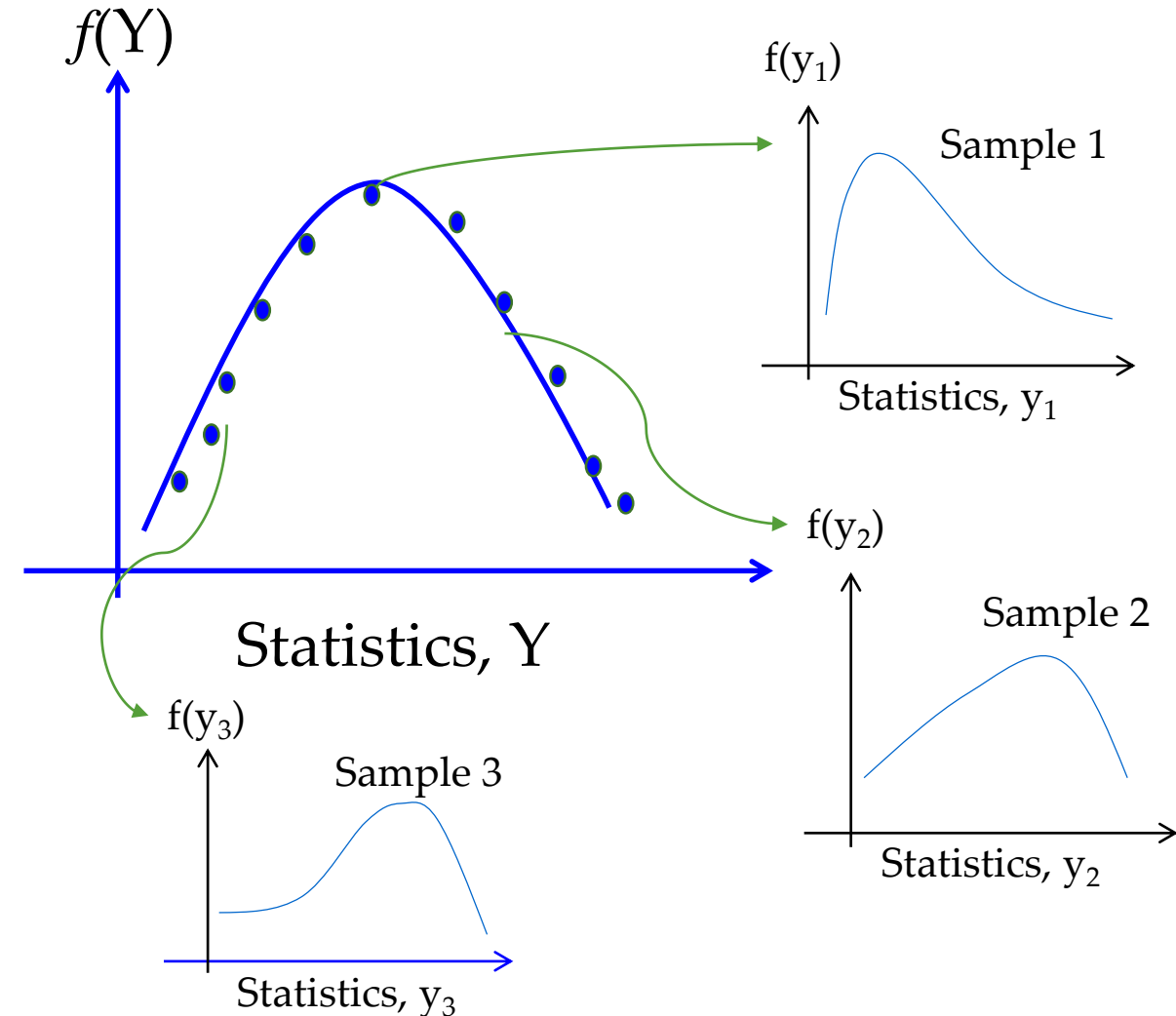


- **(Population of) Reality** : a general concept referring to the complete and defined set of possible evidences
  - Of all possible subjects (people, households, etc.) of the study area
  - Of all possible days, weeks, months, years, decades, etc.
  - Of all possible components of transportation network
- **Sample Evidence**: information of a subset of population or of a subset of possible contexts
  - Can vary by size, context and extant

# Samples have Sampling Error

- Two types of **sampling error**:
  - ✓ Sampling bias: arises when specific aspects of actual information are either missing or overlooked
  - ✓ Sampling variance: it rests on the notion of inherent variations of transportation demand resulting from dynamics of demand-supply interactions
- Sampling bias can be overcome by properly specifying study context recognizing all elements of demand generation processes.
- Sample variance cannot be eliminated, but can be minimized by considering large amount of evidence (large sample size)

# Samples have Sampling Error



- **Random sample:** Any representative set of evidences
  - Key statistics may vary with sample size, sample compositions, etc.
- **Sampling Distribution:** Distribution of sample statistics. Such as distribution of sample mean, sample variance etc.
  - Sample of evidences needs to be verified against the (population of) reality
  - To have confidence on sample evidences
- **Application of statistics (Statistical Inference) is unavoidable**

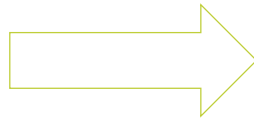
# Evidences are Collection of Measurements

- Evidences are generated by measuring different aspects of transportation demands and supply:
  - Average speed, flow, density, congestion delay etc. are measurement of network performances
  - Trip generation, trip distribution, modal split, route choice etc. are measurements of transportation demands
- We measure transportation demands by specifying variables:
  - Quantitative variables
  - Qualitative variables
- Measuring of transportation demands through variable specification has inherent measurement errors

# Construct - Measurement

- Construct: The underlying information that are to be measured is called construct
  - A single construct can be measured by specifying different variables (measurements): Each of such measurement (specified by variables) have inherent measurement error

Construct: Trip  
Generation Tendency

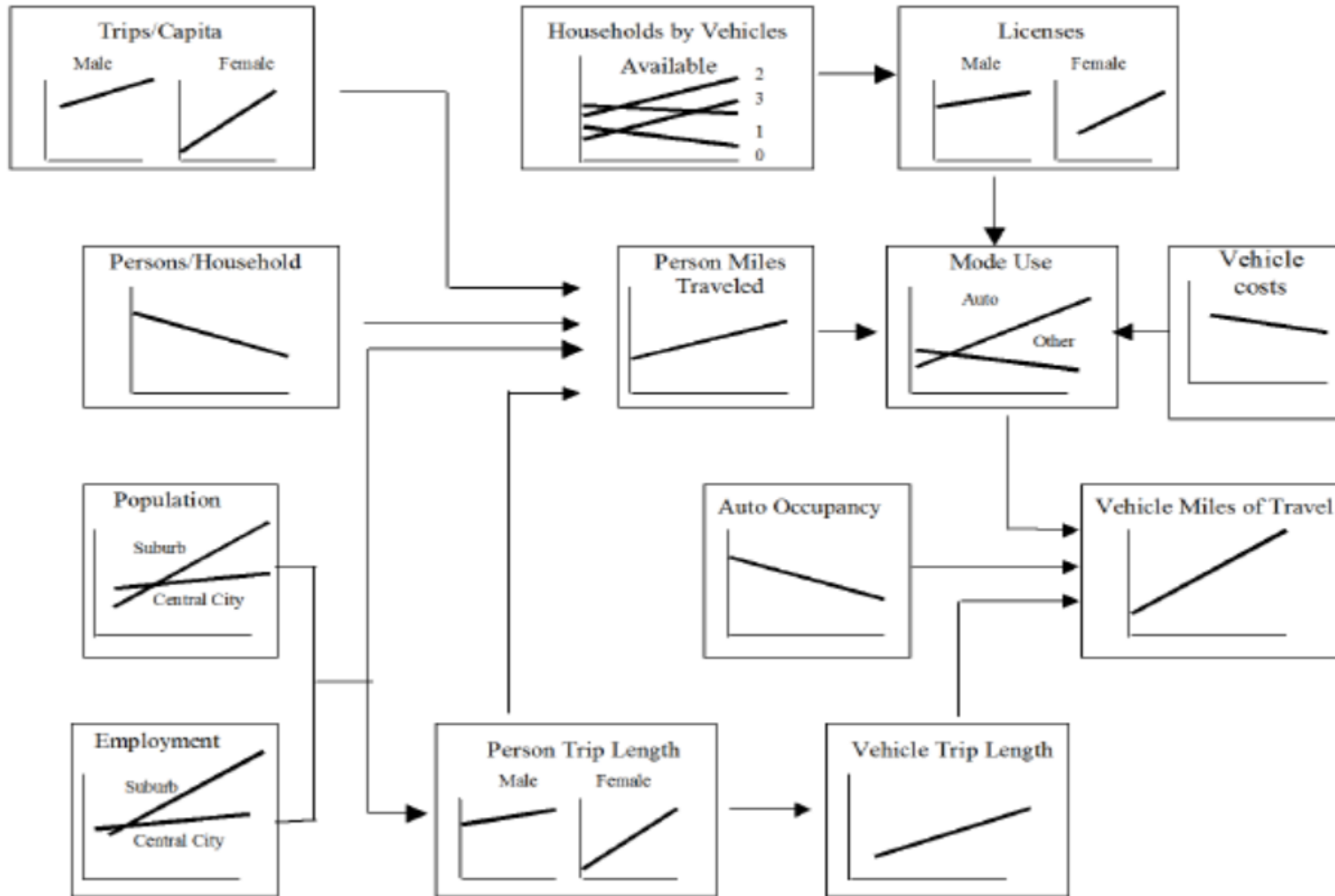


## Measurements of Trip Generation:

- Trips per day
- Trips by destination purpose
- Trips by mode of transportation
- Trips by time-of-day

- Defining construct underlying evidences (variables) of travel demand requires consideration of travel behaviour
- **Application of behavioural theory (e.g. Microeconomic) for transportation demand analysis is unavoidable**

# Complexity in Travel Demand



- Multiple variables influence a single aspect of travel demand
- All aspects of travel demand are also correlated
- Correlations among the variables complicates the errors in measurement
- Correlation among individual people and household further complicates sampling and measurement errors



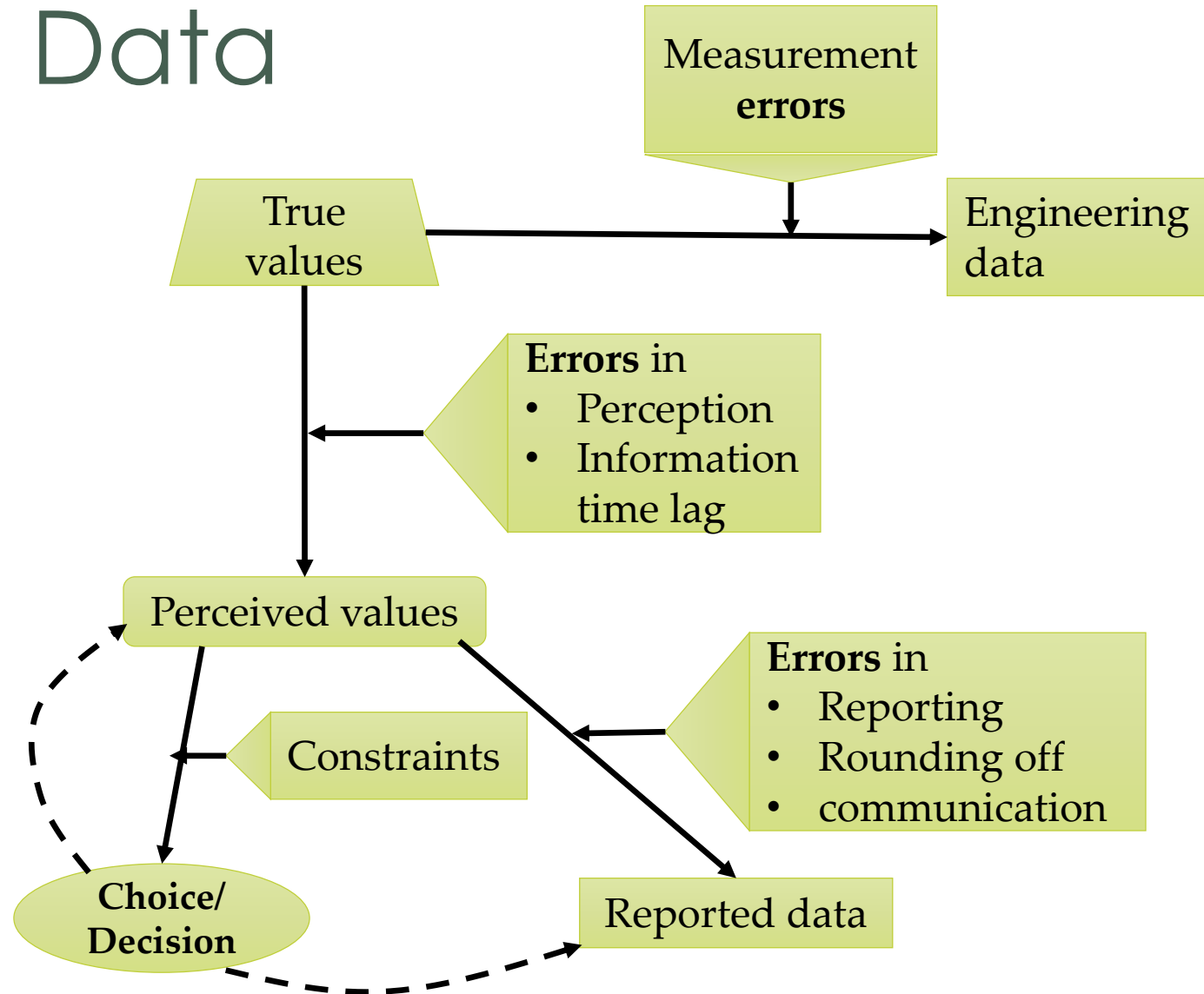
# Sampling Error, measurement Error and Evidence-based Analysis

- A collection of information (observation of variable values) does not necessarily give the appropriate evidences of transportation demands
- Appropriate evidence-based transpiration demand analysis requires
  - Observed information set (variables)
  - Definition of sample characteristics of the observed set of information
  - Identifying appropriate construct underlying the measurements presented in the observation set
  - Estimating confidence on the evidence presented in the observations set through measurements of constructs of interest
- Use of behavioural theory to specify travel demand constructs representing in a set of measured observations and estimating confidence on those measurements is **econometrics**

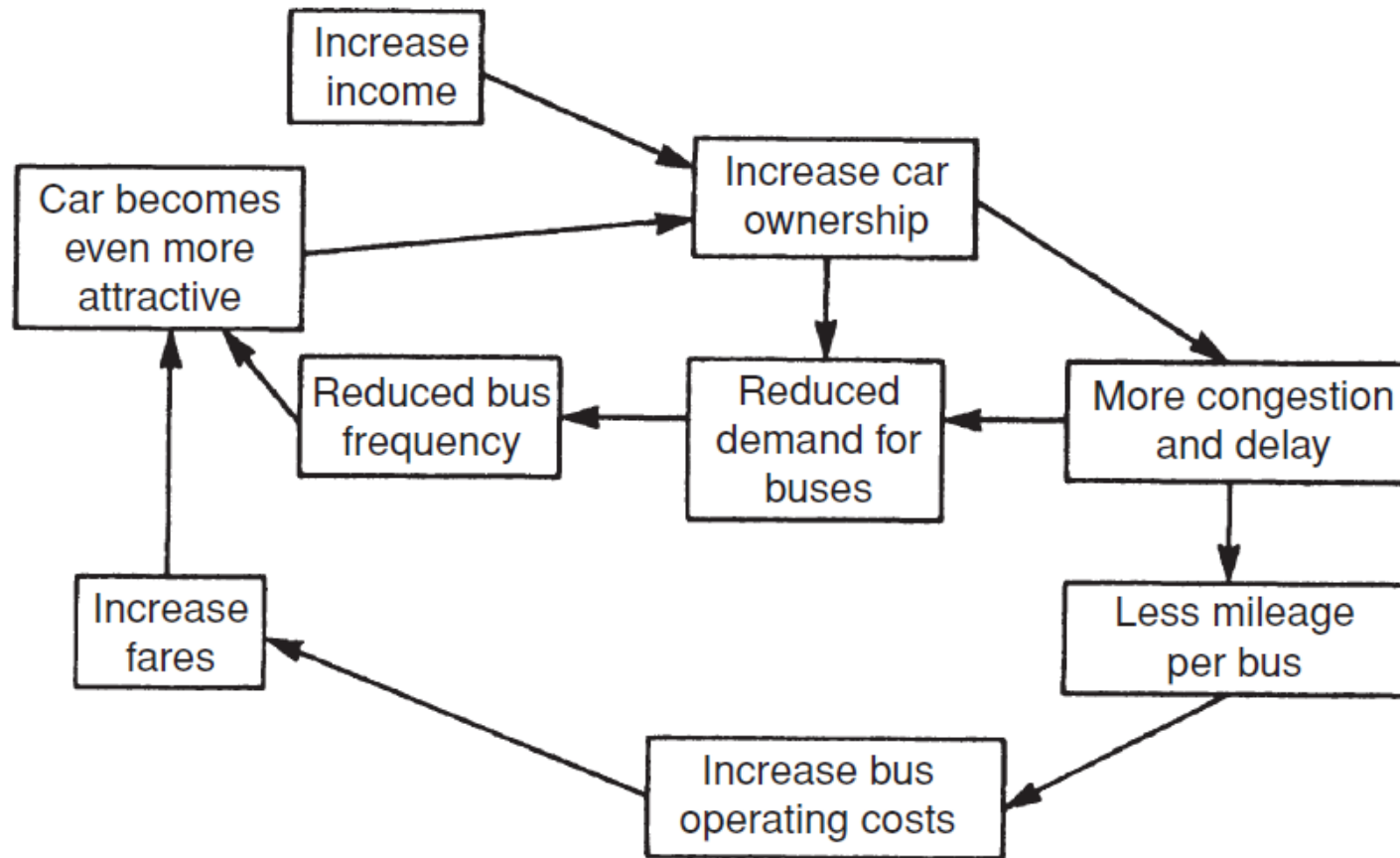
# Econometrics for Evidence-Based Analysis

- **Data (variable):** Observed set of information necessary to understand aspects of travel demand
  - Trip generation, mode choice, route choice, car ownership, home location choice etc.
- **Defining Construct:** Behavioural Theory of Travel Demand
  - Errors are nothing by unexplained variations of actual behaviour (construct) underlying measurements
  - Errors are random across the population and has specific distribution types
- **Model specification:** Specifying functional forms:
  - Systematic component of the construct that is explained by measured variables
  - Random errors components of construction with specific (appropriate) distribution
- **Model estimation:** Estimate weights (coefficients) of different variables as measured in the dataset and confidence on these estimates

# Errors in Data



# Model for Investigation



## Vicious Cycle of Car and Public Transit

- Requires definition of key element/variable/phenomena of interest: Objective variable/Dependent variable ( $y$ )
- Requires identification key Explanatory/independent variables ( $x$ )
- Requires specification of logical/mathematical relationship between dependent and independent variables
$$y = f(x, \theta)$$
- Recognize the error in model specification and variable measurement
$$y = f(x, \theta) + \varepsilon$$

# The Concept of Econometric Models

- Expressing measurement of a construction (Dependent Variable) as a function of other measurements (variables):
  - Dependent variable,  $y$ : The objective variable of interest
  - Independent variables,  $x$ : The explanatory variables
  - Weights/coefficients,  $\theta$  of  $x$  in explaining the systematic elements of measuring  $y$
  - Functional form,  $f(x, \theta)$  of the systematic explanation of  $y$
  - Random error,  $\varepsilon$  in measuring  $y$

$$y = f(x, \theta) + \varepsilon$$

- Model estimation: estimating the values as well as confidence limit of the parameters,  $\theta$ , by using a set of observed data
- Specification of variables  $y$  as well as the distribution of  $\varepsilon$  define the estimation technique

# Specification of Variables

- Possible types of variables from measurement points of view
  - Qualitative variable: Categorical/Discrete variables
    - Nominal scale: sex (male, female), urban location (CBD, suburb), employment status (employed, not employed), etc.
    - Ordinal scale: raking/ordered scale of measurement, e.g. income (low, medium, high), satisfaction (low, medium, high), reliability (low, medium, high), etc.
  - Quantitative variables: Continuous variables
    - Interval/cardinal scale: Likert scale satisfaction, numerical ranking scale etc.
    - Ratio scale: continues numbers e.g. age, travel distance, cost, etc.

# Specification of Variables

- Possible types of (dependent) variables from modelling points of view
  - Continuous variable → Continuous Econometric Model
    - Pure continuous numbers including 0 as a possible value
    - Log-transformed continuous number for positive values only
    - Other non-linear transformation: box-cox transformation
  - Discrete variable → Discrete Econometric Model
    - Binary variable (1 or 0)
    - Nominal variables with more than 2 possible values or ordinal variables can be further specified as a set of binary variables for each category

# Model Parameter Estimation

- Least-Square Estimation: Method of Moments
  - Dependent variable is a continuous variable

$$y = f(x, \theta) + \varepsilon \quad y_{prediction} = f(x, \theta) + \varepsilon$$

- For a set of observed  $y$  and corresponding  $x$  values, there will be error/residual due to the presence of  $\varepsilon$

$$E = y_{observed} - y_{prediction} = y_{observed} - f(x, \theta)$$

- Minimize the (E) for the optimum values of  $\theta$
- Least square estimation technique is unbiased, consistent and efficient.
- Microsoft excel has functions to conduct least-square estimation.
- Other software: R, Stata, LIMDEP, SAS, etc.



# Model Estimation

- Maximum likelihood estimation
  - Specify the distribution of the random error component ( $\varepsilon$ )

$$y_{\text{prediction}} = f(x, \theta) + \varepsilon$$

- Such specification makes the model ( $y_{\text{prediction}}$ ) a random variable of the same distribution of ( $\varepsilon$ ) and so, one can define the likelihood (probability) of an observed value ( $y_{\text{observed}}$ ) of a record in the dataset  $i$ ,  $L_i$   
*likelihood of an observation,  $L_i = \Pr(y_{\text{observed}})$*
- Considering that the observed dataset has  $N$  number of observation and each observations are independent, the sample likelihood becomes the multiplication of individual likelihoods

$$\text{Sample likelihood, } L = L_1 \times L_2 \times L_3 \times \dots \times L_N$$

- Maximize  $L$  (minimize  $\log$  of  $L$ ) to find out the optimum values of  $\theta$
- A non-linear optimization and standard numerical methods are available
- Max Likelihood estimation technique is unbiased, consistent and efficient
- Software: R, Stata, LIMDEP, SAS, etc.

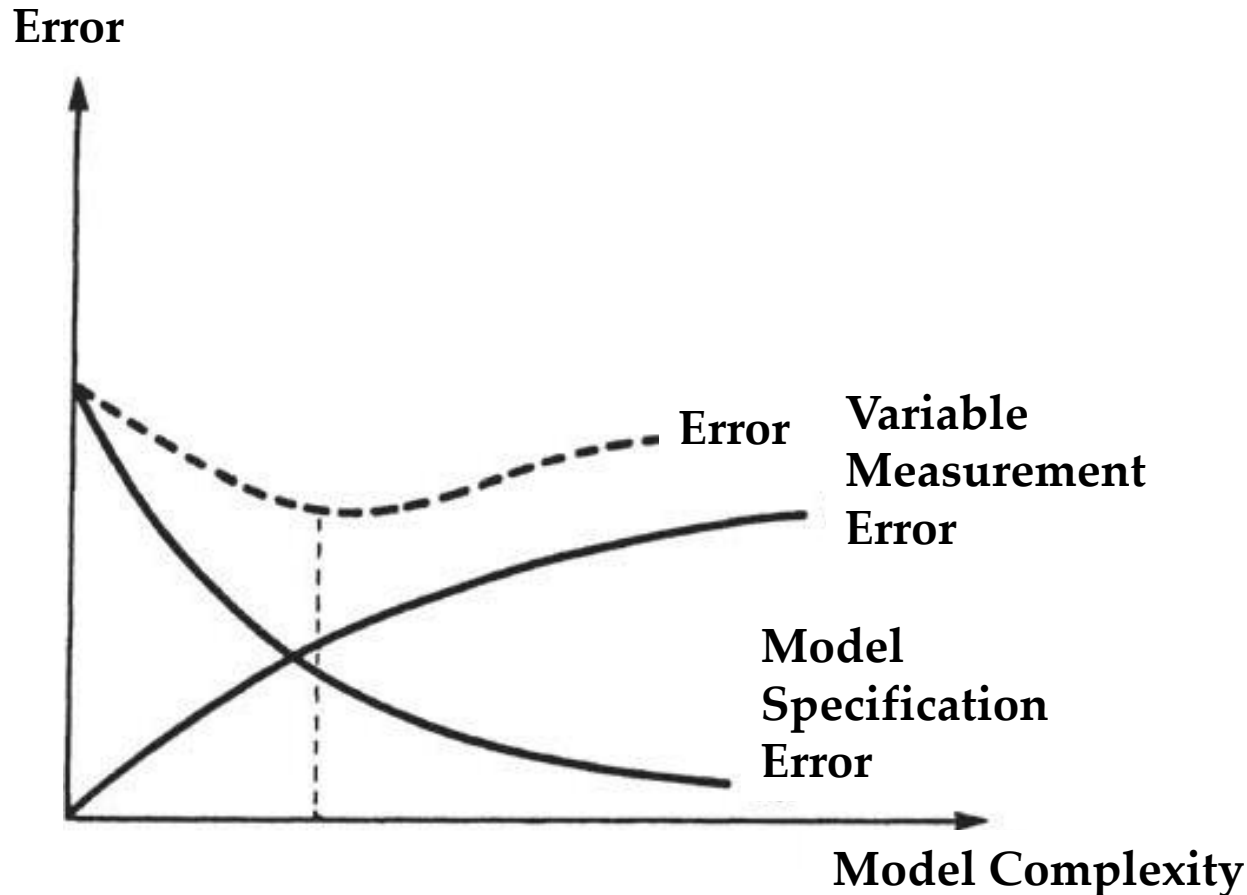
# Notes on Model Estimation

- Least-Square estimation is feasible for mostly if the dependent variable is continuous in nature
- Maximum likelihood estimation can be applied for both discrete and continuous variable models
- These estimation methods have necessary statistical properties:
  - Requires a sample of observation (dataset) with dependent as well as independent variables
  - Estimations process gives the mean value and standard errors of the parameters estimates ( $\theta$ ): So, statistical significance test can be done
  - Estimation process allows testing the goodness-of-fit of the observed values (R-squared value, Rho-squared values)

# Notes on Model Estimation

- Statistical significance of estimated parameters gives the confidence on the estimated effects of an independent variable ( $x$ ) on the dependent variable ( $y$ )
  - Ratio of estimated mean and the standard error (mean of  $\theta$  / st. err of  $\theta$ ) gives t-statistics
  - For 95% confidence limit a t value of 1.64 confirms that the corresponding estimated parameter ( $\theta$ ) value is statistically significant if the sign of  $\theta$  is known
  - For 95% confidence limit a t value of 1.96 confirms that the corresponding estimated parameter ( $\theta$ ) value is statistically significant if the sign of  $\theta$  is unknown
- A 95% confidence on estimated parameter value means, for the 95 out of 100 random sample of observation dataset, the estimated value will be very close to actual estimated value.

# Errors in Measurement and Implications



- Errors in modelling (prediction)
- Modelling complexity:
  - Increasing the number of influential variables
  - Comprehensive functional forms
- Variable measurement error
- If data are of poor quality, it may be safer to use simpler model specification

# Key Concepts of Demand Investigation

- Marginal effects
- Demand elasticity
- Substitution patterns
- Complementary versus supplementary relationships
- Income/budget effect
- Lagged effects
- Demand/Preference Heterogeneity
- Heteroskedasticity
- Choice versus Demand



# Thank You