

## Lecture 8: Matlab APP



### Class definition in Matlab

```
classdef (ClassAttributes) ClassName < SuperClass1 & SuperClass2
   properties (PropertyAttributes)
       ...
   end
   methods (MethodAttributes)
       • • •
                                                    ClassName
   end
                                                     -PropertyName
   events (EventAttributes)
                                                     -MethodName
       EventName
   end
end
```



### Class Attributes

#### • Abstract

- If specified as true, this class is an abstract class (cannot be instantiated).
- classdef (Abstract = true) ClassName

#### Sealed

If true, this class cannot be subclassed.



#### Value Class vs. Handle Class

- Value Class
- Each assignment creates a new copy of the object

```
classdef NumValue
 properties
   Number = 1
 end
end
```

- a = NumValue;
- b=a;
- a.Number = 7;
- b.Number
  - ans=1

- Handle Class
- Upon construction a reference to the object is created

```
classdef NumHandle < handle
 properties
   Number = 1
 end
```

- end
- a = NumHandle;
- b=a;
- a.Number = 7;
- b.Number
  - ans=7



## Value Class vs. Handle Class (cont.)

- When object passed into a function
  - Value object: a new copy of the object is created inside function workspace
  - Handle object: a copy of the handle (reference) is created instead of the object
- Deleting a handle object
  - Delete(NumHandle)



# Object equality

#### Value object

- Can only evaluate whether value of the objects are the same
- a = NumValue;
- b = NumValue;
- isequal(a,b)
- ans=1

#### **Handle object**

- Can check whether they are the same object as well as their value equality
- a = NumHandle;
- b = a;
- a == b (same object?)
  - ans=1;
- isequal(a,b) (same value?)
  - ans=1;

- a = NumHandle;
- b = NumHandle;
- a == b
  - ans=0;
- isequal(a,b)
  - ans=1;



### Class Members Access

- public Unrestricted access
- protected Access from methods in class or subclasses
- private Access by class methods only (not from subclasses)
- List classes (and their subclasses) have access to this member
  - (Access = {?ClassName1,?ClassName2,...})



## Property Attributes

- Read and write access
  - GetAccess
  - SetAccess
    - properties(GetAccess = 'public', SetAccess = 'private')
    - % public read access, but private write access.
    - end
    - SetAccess = immutable: set during construction, cannot be changed afterwards
- Constant

```
properties(Constant = true)
     DAYS_PER_YEAR = 365;
end
```

- Dependent
  - depend on other values
  - calculated only when needed.
  - i.e. area of a square depends on the width property



### Class Constructor Method

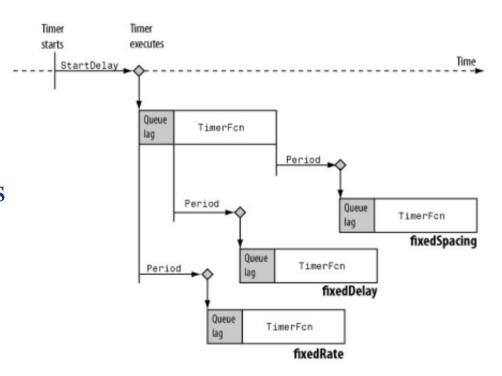
- There is a default class constructor without input arguments
- We can define class constructor that overrides the default one
- Method with the same name as the class name

```
classdef ConstructorDesign < BaseClass1
methods
  function obj = ConstructorDesign(a,b,c)
  end
end</pre>
```



### Timer Class

- t = timer;
- Properties
  - 'ExecutionMode'
  - 'Period': Time between timer functions
  - 'TimerFcn': Function handle
- t.TimerFcn=@callback;
- Function callback(hObj,src,event)





# Demo: Traffic Light





## Example: Information system for restaurants

• The owner of restaurant A would like to improve service efficiency



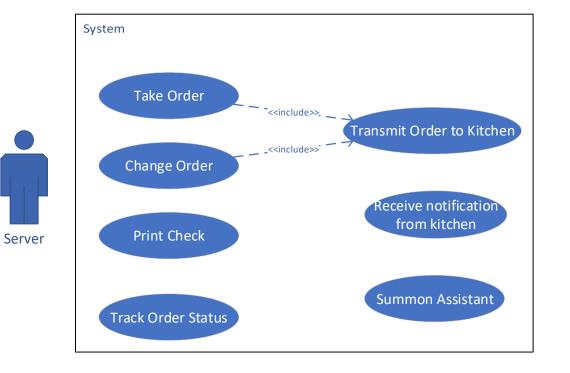


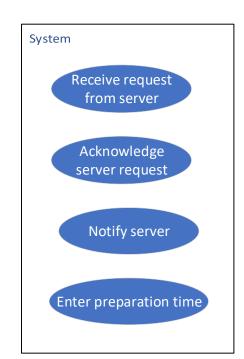




# Discover system requirements (cont.)

• System requirements as use cases

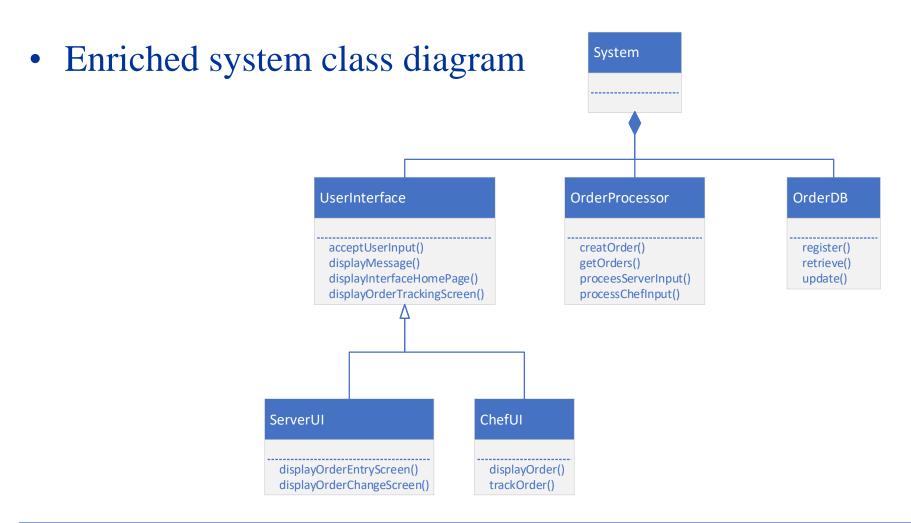




Chef



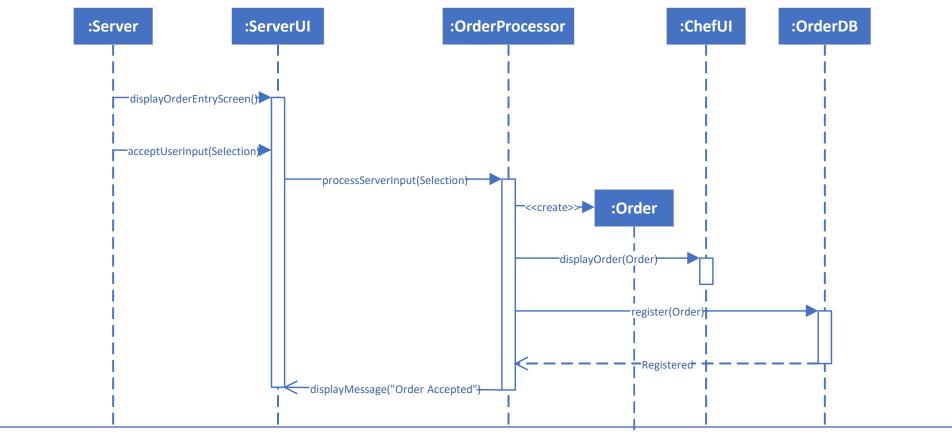
## Discover system requirements (cont.)





## Identify interactions

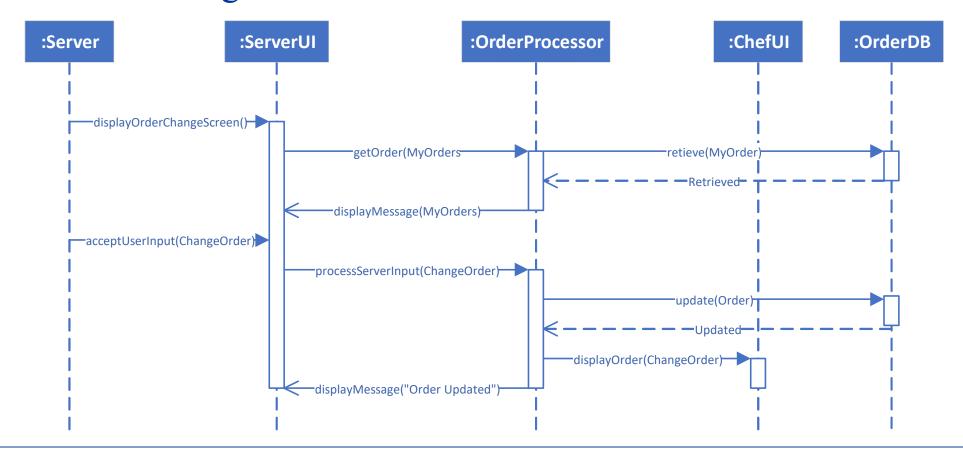
• Use case "Take an order"





## Identify interactions (cont.)

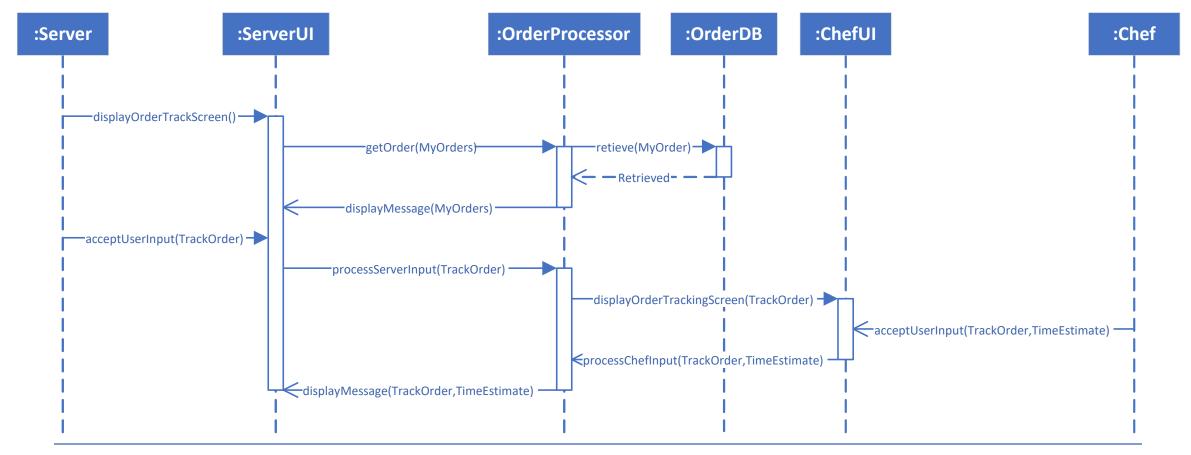
• Use case "Change an order"





## Identify interactions (cont.)

• Use case "Track an order"





## Why do we need models?

#### Prediction

- We know the low-level mechanisms but we want to understand how they affect higher-level behaviors
- Use simulation instead of testing on the real system

#### Explain the data

 Make assumptions and use our knowledge to explain mechanisms that we don't understand

#### Classification

- i.e. definitions, machine learning algorithms



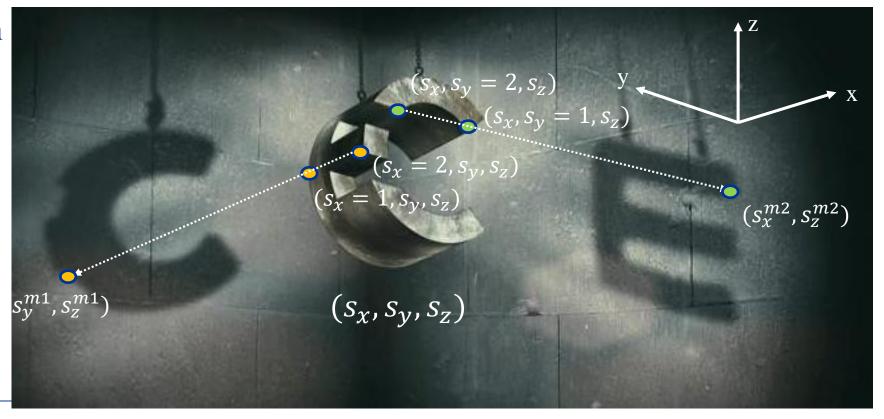
### What are models?

- A system: (*S*, *I*, *T*, *O*)
  - S: States  $s_1, s_2 \dots s_n$
  - *I*: Inputs (could be ∅)
  - T: Transitions  $S \times I \times S$
  - *O*: Observations  $f(S_o)$ ,  $S_o \subseteq S$

- Model of the system  $(S^m, I^m, T^m)$ 
  - $S^m$ : Abstraction/interpolation of S
    - Much fewer state variables
  - $I^{\mathrm{m}}$ : abstraction of I (could be  $\emptyset$ )
  - $T^m$ : Transitions  $S^m \times I^m \times S^m$

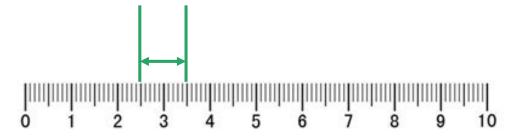
### Abstraction – removal of state variables

- States  $(s_x, s_y, s_z)$  are abstracted to  $(s_y^{m1}, s_z^{m1})$ 
  - $\left( S_{\chi}, S_{y}, S_{z} \right) \rightarrow \left( S_{y}^{m1}, S_{z}^{m1} \right)$
- Loss of information



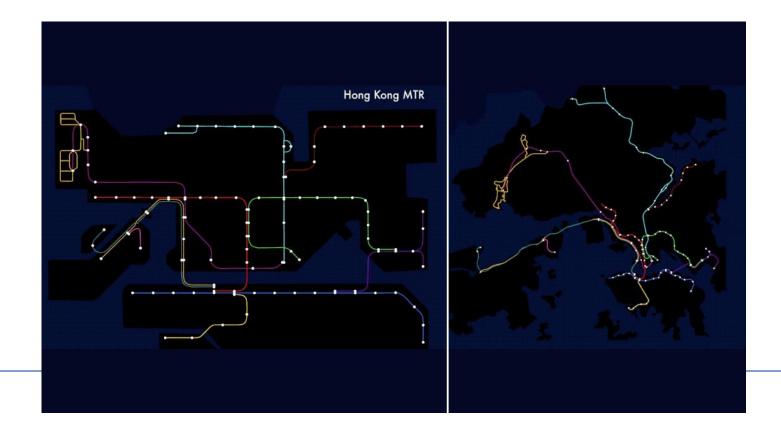
### Abstraction: Approximation of state variable values

- Irrational numbers
  - $-\pi \approx 3.1415$
  - $-\sqrt{2}\approx 1.414$
- Approximation is another way of abstraction



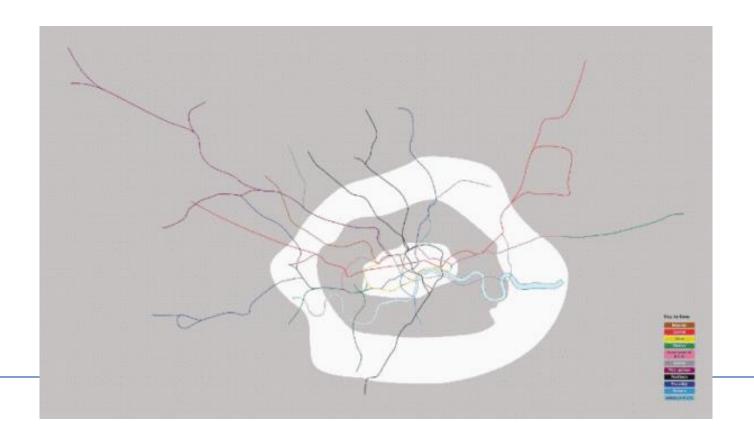
## Interpolation: extracting interpretable information

- Locational information -> topological information
- $S^m = f(S_p), S_p \subseteq S$





## More Interpolation: London MTR





# What is considered as a "good" model?

- Accuracy
  - All models are wrong!
  - Error accumulates over time
  - Initial condition of the model cannot be determined due to limited observability
- Generality
  - The capability to explain not only training data, but also testing data
- Identifiability
  - Model parameters can be identified from data
- Interpretability
  - $-S^m$  are meaningful and interpretable by human

### Newton vs. Einstein

• Newtonian physics is suitable for macro level objects at low speed

$$\bullet \ L = L_0 \sqrt{1 - \frac{v^2}{c^2}}$$

• A model can only be "good" within the context of its designated application

• The definition of "goodness" is changing over time



# Modeling methodologies

- Bottom-up modeling
  - "White-box" model
  - Using first principles
  - Pros:
    - Interpretable
    - Convincing
  - Cons:
    - State space explosion
    - Difficult to be general
    - Low identifiability



- Data driven models (i.e. Neural networks)
  - "Black-box" model
  - From observable data
  - Pros:
    - No need to know domain knowledge
  - Cons:
    - Large and uninterpretable  $S^m$
    - Depends highly on the quality and quantity of data