

Fall 2010 Graduate Course on
Dynamic Learning

Chapter 1: Introduction

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Outline

- Motivating examples
 - AI: Vision, Language, Robotics
 - BI: Brain and Cognitive Science
- Dynamical systems
 - Definition
 - Types
- Learning in dynamical systems
 - Sequential/online
 - Learning and action interleaved

Vision: Target Tracking



- Multiple people tracking

(<http://www.youtube.com/watch?v=sk9WY79w8oM&feature=related>)

Speech Recognition



- The virtual personal assistant on your phone (Siri)
(<http://www.youtube.com/watch?v=MpjpVAB06O4>)

Language Generation



Watch from 2:00 min

Multimodal conversational agents

(<http://www.techfak.uni-bielefeld.de/ags/wbski/media/MAXMTV.MPG>)

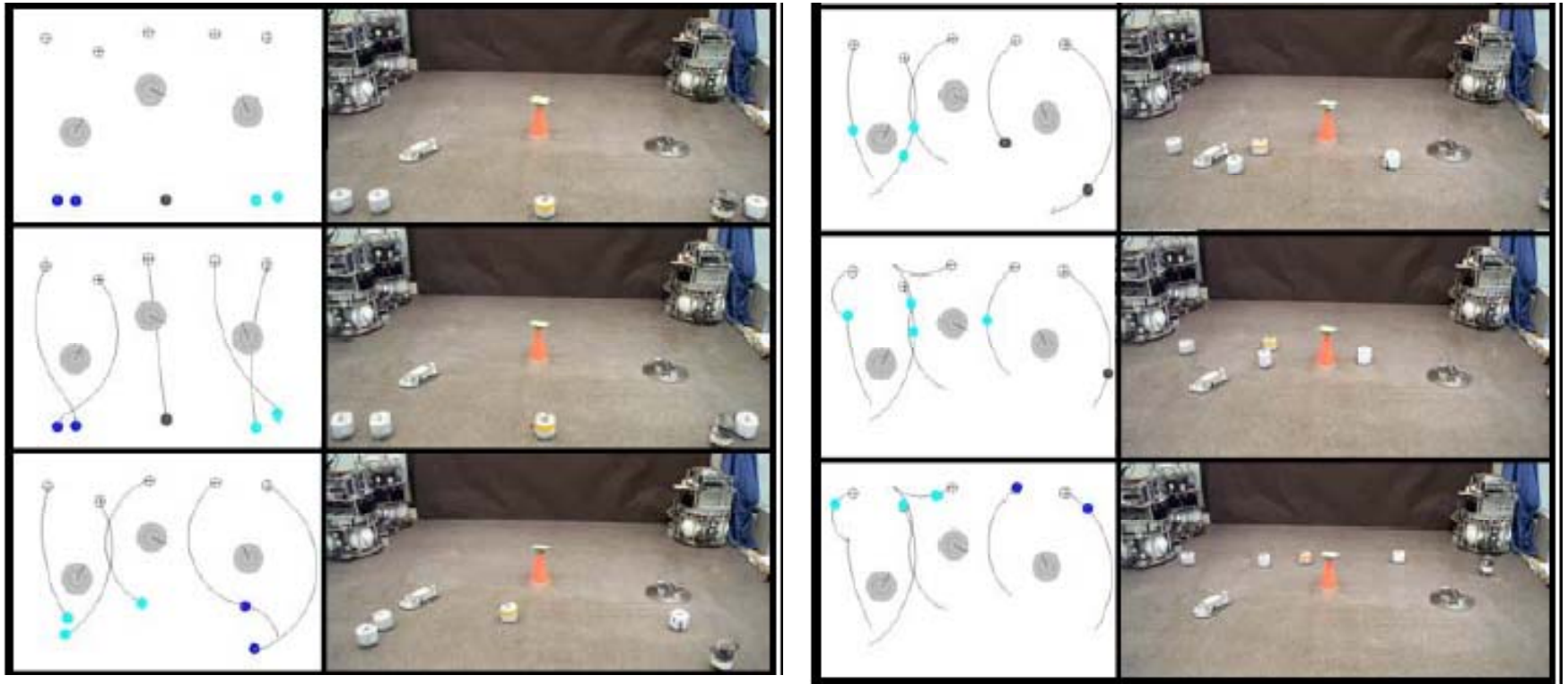
Graphics: Motion Planning



Watch from 3:00

- Real-time scalable motion planning for crowds (<http://www.youtube.com/watch?v=ifimWFs5-hc>)

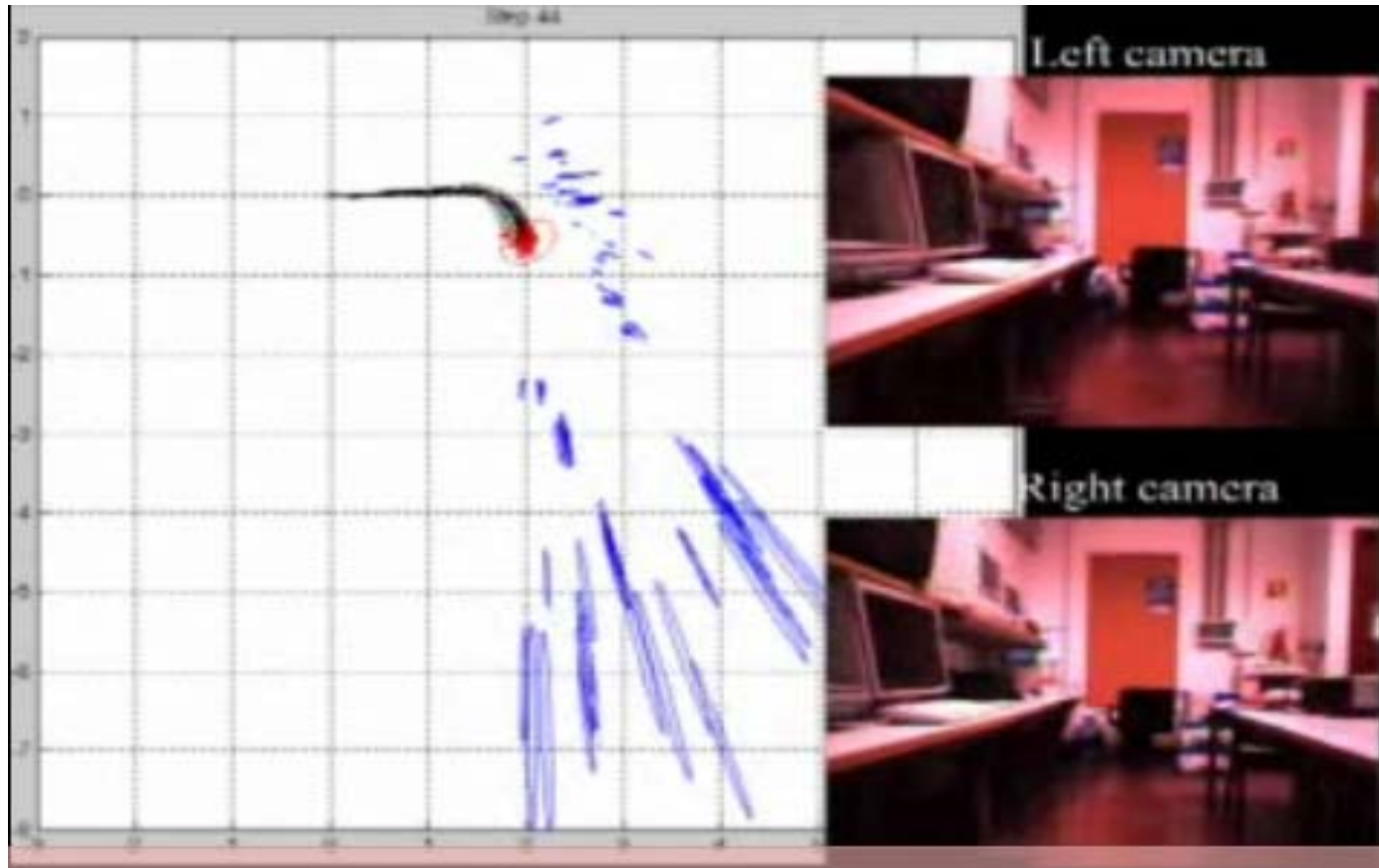
Robotics: Path/Motion Planning



Motion planning for multiple mobile robots

Example experiment on the Micro-Autonomous RoverS (MARS) test-platform involving 5 robots and 3 obstacles. (C.M. Clark *et al.*, *IEEE International Conference on Robotics and Automation*, 2003)

Robotics: Localization

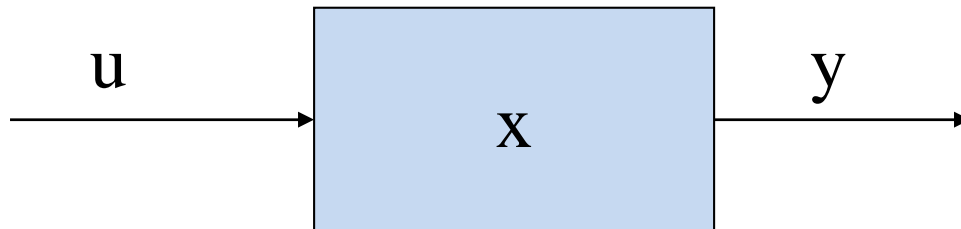


- Real-time localization in indoor with stereo vision (<http://www.youtube.com/watch?v=m3L8OfbTXH0&NR=1>)

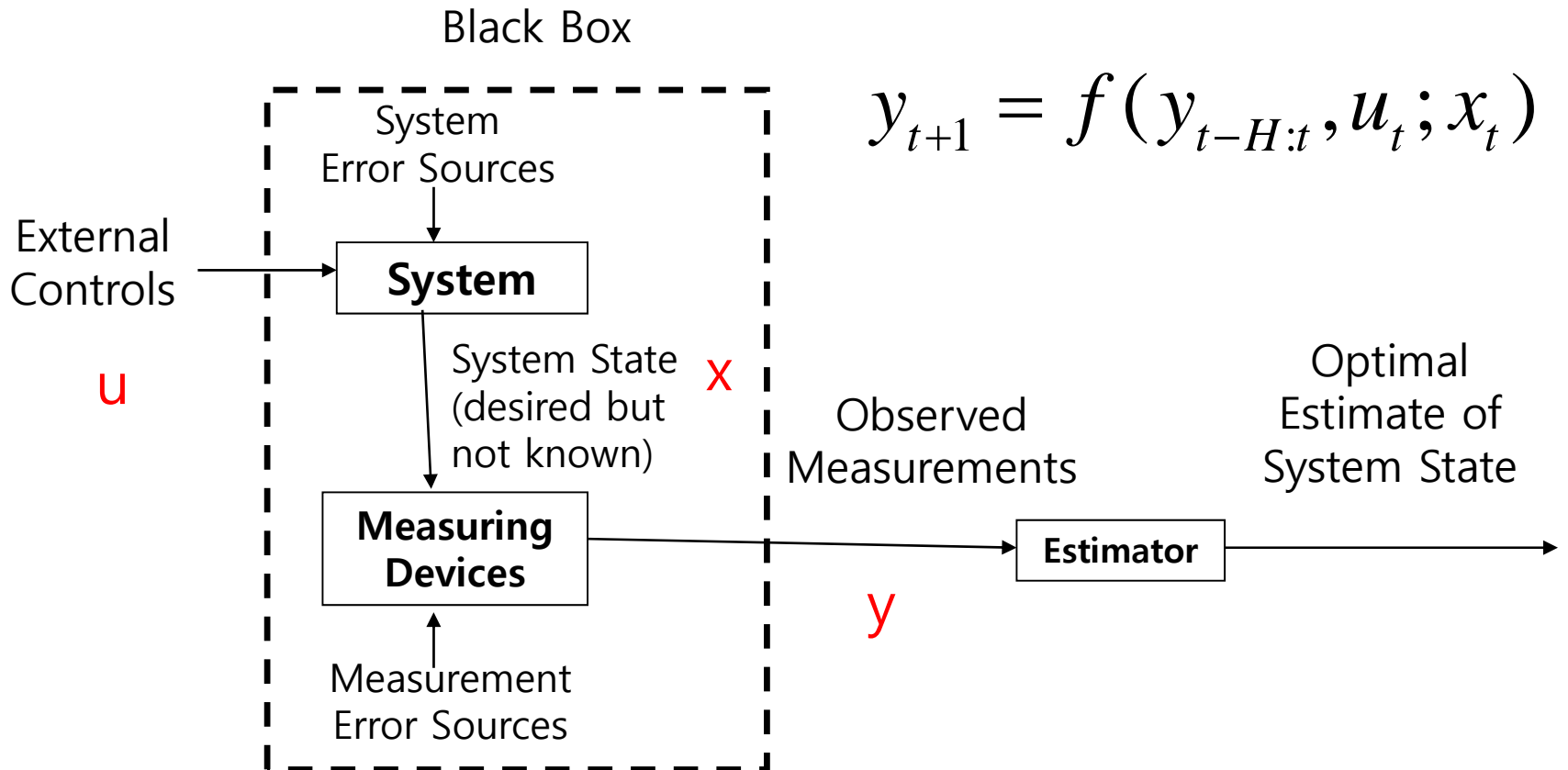
Dynamical Systems

Dynamical systems

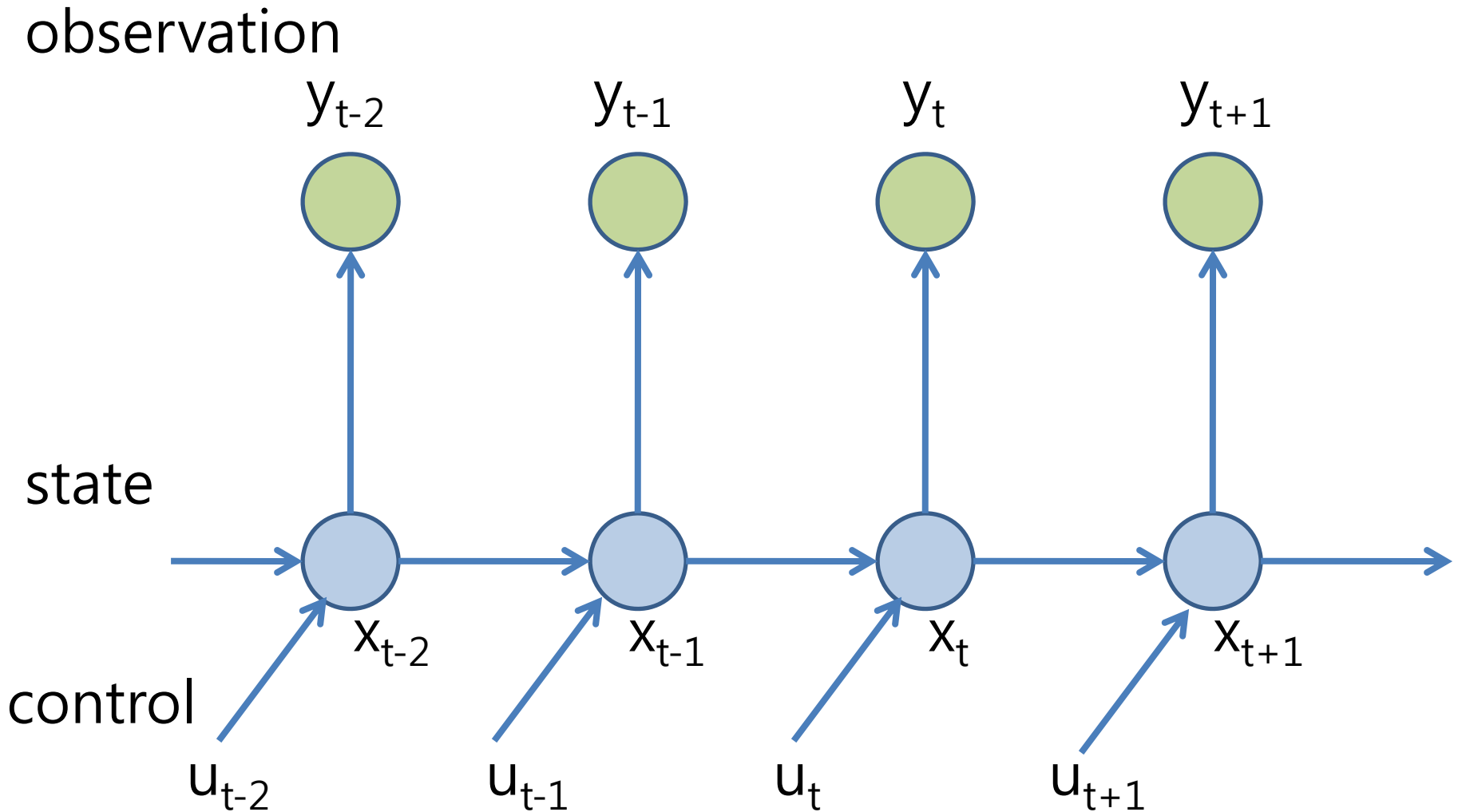
- **System**: a (physical) entity which can be manipulated with actions, called **inputs** (u) and that, as a consequence of the actions, gives a measurable reaction, called **output** (y)
- **Dynamic**: the system **changes over time**; in general, the output does not only depend on the input, but also on the current "**state**" of the system (x), i.e. on the system **history**



Dynamical system



Graphical model



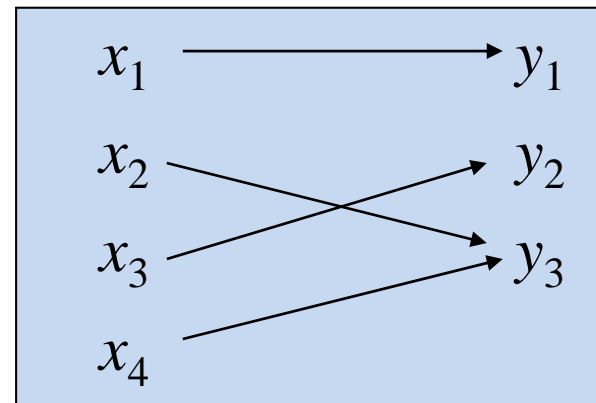
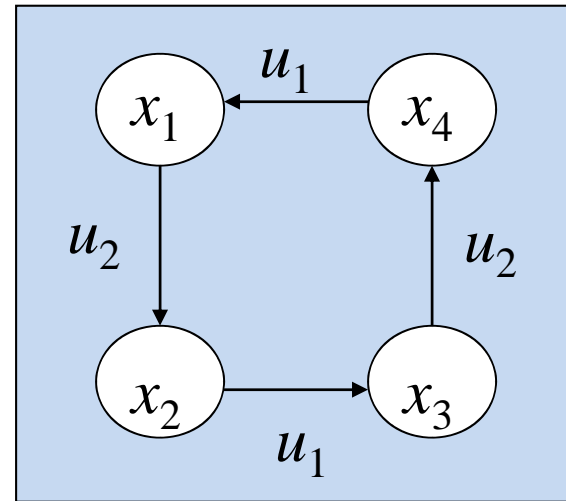
Modeling a dynamical system

- Two ingredients:
 - State transition function

$$x_{t+1} = f(x_t, u_t)$$

- Observation function

$$y_t = g(x_t)$$



Main classes of dynamical systems

- Continuous / discrete
- Linear / nonlinear
- Time invariant / variant systems
- Single / multiple input / outputs
- Deterministic / stochastic

Discrete and continuous systems

- **Discrete**: the time set is the set of integer numbers ($t=1,2,\dots,k,\dots$). The system is typically modeled with difference equations

$$x_{k+1} = f(x_k, u_k),$$

$$y_k = h(x_k)$$

- **Continuous**: the time set is the set of non-negative real numbers. The system is typically modeled with differential equations

$$\frac{dx}{dt} = f(x_t, u), \quad x_{t_0} = x_0, \quad y_t = h(x_t)$$

Learning in Dynamical Systems

Dynamic Learning

- Target system
 - Changing environment
 - Moving target (can be nonstationary)
 - Sequential decision making
 - Action (control) selection
 - Prediction/forecasting
- Learning system
 - Sequential observation of data
 - Online learning (vs. batch learning)
 - Interactive learning (action interleaved)
 - Sequential/recursive estimation
 - Short-term vs. long-term objectives

Course Objectives

- Review existing models for **dynamic learning**
- Develop a **sequential Bayesian** framework for dynamic learning
- Emphasis on Monte Carlo filtering methods, such as **particle filters**
- Establish the connection between **particle filtering and evolutionary optimization**
- Construct graphical models for dynamic Bayesian learning from **video (language + vision) stream** data

Project

- Given
 - A sequence of T image-text pairs of $Y(t) = (V(t), L(t))$, $t=1, \dots, T$
 - $V(t)$: a vector of visual words, $L(t)$: a vector of linguistic words
 - E.g., from a 20-minute episode of *Friends*
- Construct
 - A dynamic system that learns to estimate the (mental memory) states to generate the future image-text sequences from a historical context of size H .

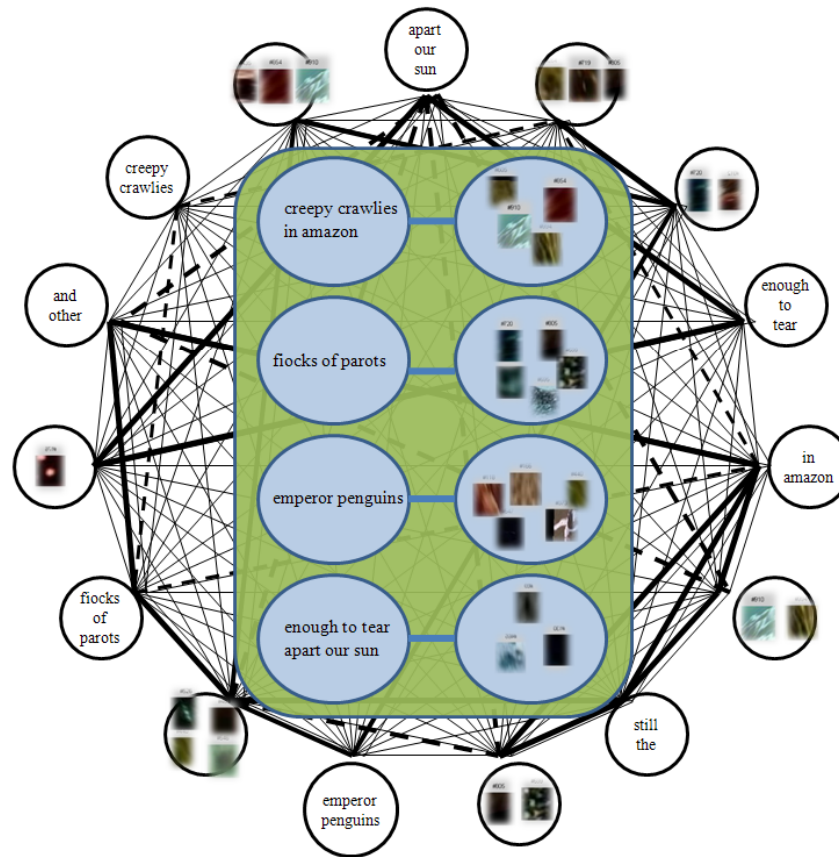
$$y_{t+1} = f(y_{t-H:t}, u_t; x_t)$$

- To demonstrate
 - **Visual storytelling**: Given a sequence of pairs of $(V(h), L(h))$, $h = t, t-1, t-2, \dots, t-H$, generate $(V(r), L(r))$ for $r = t+1, t+2, t+3, \dots, t+R$
 - L2V translation (**mental imagery**): Given a series of texts, generate a series of images.
 - V2L translation (**scene description**): Given a series of images, generate a series of texts.

Dynamic Learning Memory



Pre-Processing



Project: Visual Storytelling

observation

