WebWatcher

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Question

- WebWatcher가 학습을 하는 방법에 대해 나열하시오.

Outline

- What is the WebWatcher?
- Structure of WebWatcher
- Operation of WebWatcher
- Learning in WebWatcher
- Experiment
- Related Work
- Summary and Future Research

What is the WebWatcher?

- Browsing the WWW is like visiting a museum
- Accompanies the user as he or she browses the Web like a museum tour guide
- User can communicate with the system and give feedback
- Over time, learn to acquire greater expertise for previous visiting parts and interest types

Difference from search engine

- Search engine
  - Require specific words as keywords
  - Match keywords in the target Web page
  - Documents are not designed as hypertext
- WebWatcher
  - Learn that “machine learning” matches a hyperlink such as “neural networks”
  - Self-improve

Structure of WebWatcher

- WebWatcher is implemented as a server
- Acts much like a proxy
- Return a page to the user by three modifications
  - Command list is added to the top
  - A original URL is replaced by a new URL

    Ex) http://webwatcher.learning.cs.cmu.edu:8080/cgi-bin/agent-welcome.pl?http://www.cs.cmu.edu/Web/FrontDoor.html
  - Highlight the most promising links

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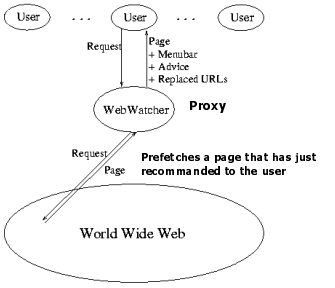
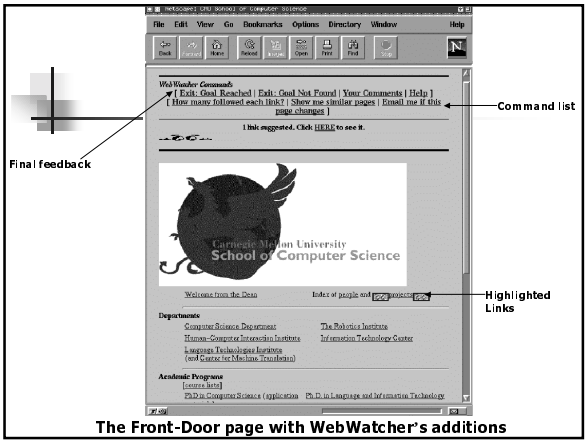


Operation of WebWatcher

- Invoked by clicking on the hyperlink "The WebWatcher Tour Guide"
- Leads us to a page of our current interest
- Accompanies us from the additions
  - WebWatcher Commands : communication method
  - Highlighted hyperlinks : suggestion directions
- To end the tour, two options in the command list
  - Exit : Goal reached
  - Exit : Goal not found



The Front-Door page with WebWatcher's additions

Learning in WebWatcher

- LinkQuality
  - The probability that a user will select *Link* given the current *Page* and *Interest*
  - $LinkQuality : Page \times Interest \times Link \rightarrow [0,1]$
- Three approaches to learning this target function
  - Learning from Previous Tours
  - Reinforce learning
  - Combined method

Learning from Previous Tours

- Annotating each hyperlink with the interest on previous tours.
- Compare current user's interest with descriptions of hyperlinks
- Interests and hyperlink description are represented by very high-dimensional feature vectors
- Elements of a vector are calculated using the TFIDF heuristic
- Similarity is calculated as the cosine between vectors

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- The value of *LinkQuality* for each hyperlink is the average similarity of the *k* keyword sets for this hyperlink
- Suggested if *LinkQuality* is above a threshold
- Maximum number of Suggested hyperlink is three

Keyword Vectoring

- Removes the suffix of the words
- Filter all commonly-used words
- Each word is weighted by TFIDF measure
- Term Frequency Inverse Document Frequency
  - $V_i = \Pi_c \times \text{Freq}(\text{Word}_i) \times [\log_2(n) - \log_2(\text{DocFreq}(\text{word}_i))]$
- Collection of documents is the set of all weighted keyword vectors

Learning from Hypertext Structure

- Augments a given hyperlink using words encountered in pages downstream of it
- Reinforcement Learning
  - Learns control strategies that select optimal actions in certain settings
  - The objective is to find paths through the Web which maximize the amount of relevant information

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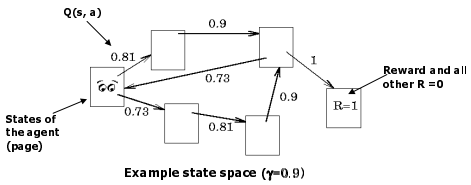
- Discounted sum of future rewards

$Q(s_t, a) = \sum_{i=0}^{\infty} \gamma^i \cdot R(s_{t+1+i})$ **The goodness of an action a**

S_t: state in at time t γ: discount factor($0 \le \gamma < 1$)

R(s): reward a : action



Example state space (γ=0.9)

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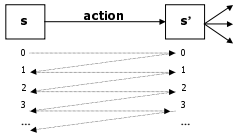
- Under certain conditions, Q function can be iteratively approximated by updating estimate for Q(s, a) repeatedly

$$Q_{n+1}(s, a) = R(s') + \gamma \max_{a' \in actions\_in\_s'} [Q_n(s', a')]$$



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- Reinforcement Learning and Hypertext
  - Interest "*intelligent*"
    - $R_{intelligent}(s)$: the TFIDF value of "*intelligent*" for page s
    - $Q_{intelligent}(s, a)$: the sum of discounted TFIDF values of "*intelligent*" over the optimal tour beginning with a
  - WebWatcher uses a separate reward function $R_w(s)$ and learns a distinct $Q_w(s, a)$ for every word w
  - Problem
    - WebWatcher cannot expect that users will always stick to pages it has already seen

⇨ **Distance-weighted 3-nearest neighbor function approximator**

Experiment

	Accuracy
Random	31.3%
Popularity	41.9%
Match	40.5%
Annotate	42.2%
RL	44.6%
Combine	48.9%

Random : suggests hyperlinks at random from current page

Popularity: followed most frequently in the past

Match: TFIDF-cosine similarity between underlined text and user's interest

Annotate: Learning from Previous Tours

RL: Reinforcement learning

Combine: combines above all using logistic regression

Experiment

	Accuracy
Random	22.4%
Annotate	42.9%
Human	47.5%

Comparison to human performance on the Front-Door page

Related Work

- Letizia
- Syskill and Webert
  - Manually constructed index page
  - User can rate hyperlinks off page
  - System use the ratings to learn
- Lira
  - Works in an offline setting

Conclusion and Future Research

- WebWatcher is a Self-improving tour guide agents
- Provide helpful advice to many users
- But there are some topics for future research
  - Personalized WebWatcher
  - Combining user-specific and Web locale-specific learning
  - Richer dialogs with users
  - New machine learning algorithms
  - Intelligent distributed hyperlinks