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1 Begriffe

1.1 Online-Probleme

Online-Probleme sind Probleme, bei denen die Eingabe zu Beginn nicht vollständig bekannt sind, sondern laufend hinzukommen. Sie sind eine wichtige Klasse von Problemen und treten beispielsweise bei der Planung von (CPU-)Jobs / Scheduling auf.

Definition: Online Maximierungs-Problem

Online-Probleme bestehen aus Menge von Eingaben \mathcal{I} , Eingabe $I \in \mathcal{I}$ mit $I = (x_1, \dots, x_n)$, sowie einer (*gain*)-Funktion. Zu jeder Eingabe kann eine Menge von Ausgaben mit jeweils $O = (y_1, \dots, y_n)$ zugeordnet werden. $gain(I, O)$ ordnet dabei zu jeder Eingabe und passender Ausgabe eine positive, reelle Zahl zu. Für jede Eingabe I nennen wir diejenige Lösung O eine *optimale Lösung von I* , für welche $gain(I, O)$ maximal ist. Wir schreiben diese als $Opt(I)$

Definition: Online-Algorithmus mit *Advice*

1.2 Advice-Komplexität

2 Das einfache Online-Rucksackproblem

2.1 Greedy-Ansatz

$1 - \beta$

2.2 Untere Grenze für Optimalität

$n-1$ advice-bits

2.3 AONE - 2-competitive

2.4 Grenzen für $\log(n-1)$ Advice bits

$2 - \epsilon$

2.5 SLOG

3 Vergleich mit randomisierten Algorithmen

3.1 1 Random bit: RONE - 4-competitive

3.2 1 Random bit: RONE' - 2-competitive

3.3 Mehr Random-bits: keine Verbesserung

4 Ausblick

4.1 Rucksackproblem mit Überfüllen

4.2 Das gewichtete Online-Rucksackproblem

5 todo:

- graph von bits / n, sprung-marken

6 Online-Knapsack-Problem

```
1 Constants =  
2   SCALE: 300  
3   KNAPSACK_SIZE: 1  
4  
5 roundValue = (value) -> Math.round(value*100)/100
```

6.1 Knapsack

The Knapsack will contain the objects and can return its gain (= the total of items' value)

```
1 Knapsack = class  
2   constructor: ->  
3     @size = Constants.KNAPSACK_SIZE  
4     @dep = new Tracker.Dependency  
5     @reset()  
6  
7   reset: ->  
8  
9     @items = []  
10    @dep.changed()  
11  
12    fits: (item) -> @gain() + item.value <= Constants.KNAPSACK_SIZE  
13  
14    addItem: (item) ->  
15      if @fits item  
16        @items.push item  
17        @dep.changed()  
18    gain: ->  
19      roundValue _.reduce @getItems(), ((total, item) -> total+item.  
20        value), 0  
21    getItems: ->  
22      @dep.depend()  
23      @items  
24  
25    experiments = []  
26  
27    Algorithm = class  
28      constructor: ->  
29        @adviceBits = new ReactiveVar  
30        @act = new ReactiveVar  
31        @_knapsack = new Knapsack
```

```

32     knapsack: -> @_knapsack
33     askOracle: (items) ->
34         if @oracle?
35             @adviceBits.set @oracle items
36             delete item.isPartOfSolution for item in items
37     readAdviceBit: (index) ->
38         @adviceBits.get()[index]
39     reset: ->
40         @_knapsack.reset()
41         @adviceBits.set null
42         @act.set null
43     handle: (item) ->
44         if @decide item
45             @_knapsack.addItem item
46             yes
47         else
48             no
49     doAct: (like) -> @act.set like
50     acts: (like) -> @act.get() is like

```

Lets start with the greedy aproach. Here, we just take every item we get, if it fits:

```

1 decideGreedy = (item) -> if @knapsack().fits item then yes else no

```

and we define an algorithm with it:

```

1 Greedy = class extends Algorithm
2     decide: decideGreedy

```

The gain of this algorithm is at least $1-\beta$, where β is the size of the item with the highest value (weight). Lets do some experiments with it:

```

1 experiments.push
2     name: -> "Greedy G"
3     description: -> "G archieves at least 1-beta, where beta is here
4         #{@beta}"
5     beta: 0.5
6     Algorithm: Greedy
7
8 experiments.push
9     name: -> "Greedy G"
10    description: -> "G archieves at least 1-beta, where beta is here
11        #{@beta}"
12    beta: 0.2
13    Algorithm: Greedy
14
15 experiments.push
16     name: -> "Greedy G"
17     description: -> "G archieves at least 1-beta, where beta is here
18         #{@beta}"
19     beta: 0.8
20     Algorithm: Greedy

```

6.2 Advice bits

Imaging you had an oracle, that would know all items that will come. How many bits of information from this oracle would you need to get an optimal solution? And for a given amount of these advice bits, how good would your algorithm perform?

Let's start with the first question.

Consider an algorithm with an oracle, that would give us a bit for every item coming with

- value 1 if the item is part of the solution
- value 0 if the item does not belong to the solution

We now define an algorithm for that.

Note: The items are prepared in a way, that some are already marked as solution. That makes it easier to define the oracle here:

```

1 TotalInformation = class extends Algorithm
2   oracle: (items) ->
3     bits = []
4     for item in items
5       bits[item.index] = if item.isPartOfSolution then 1 else 0
6     # we do not need the last (n-1)
7     bits.pop()
8     return bits

```

The decision is now easy. If we have a bit (yes / no), we use it:

```

1   decide: (item) ->
2     adviceBit = @readAdviceBit item.index
3     if adviceBit? then adviceBit else yes

```

Lets do an experiment with it:

```

1 experiments.push
2   name: -> "Total Information"
3   beta: 0.4
4   Algorithm: TotalInformation

```

6.3 1 Advice bit

What's the best gain if we had only 1 advice bit?

Let's do an experiment where we have an oracle that gives us one bit:

```

1 AONE = class extends Algorithm
2   oracle: (allItems) -> [ _.some allItems, (item) -> item.value >
3     0.5 ] # array with one bit

```

The bit tells us:

- 1: There exists an item with a size > 0.5
- 0: There is no such item

If the bit is 0, the algorithm acts greedy (like before). If the bit is 1, the algorithm waits until the item with size > 0.5 appears and will start acting greedyly:

```

1   decide: (item)->
2     adviceBit = @readAdviceBit item.index
3     if adviceBit? # existence
4       if adviceBit is false then @doAct "greedy" else @doAct "
5         wait"
6       if @acts "greedy" then decideGreedy.call @, item else @wait
7         item

```

```

6
7     wait: (item) ->
8         if item?.value > 0.5
9             @doAct "greedy"
10            decideGreedy.call @, item
11        else
12            no

```

We do an experiment with a max size of one item of 0.55:

```

1 experiments.push
2   name: "AONE - with one advice bit"
3   description: "AONE is 2-competitive"
4   beta: 0.55
5   Algorithm: AONE

```

6.4 Random Online-Algorithms

Obviously in real online-problems, we do not have an omniscient oracle. But we can use the idea of the oracle and just guess the advice bits *randomly*.

We can then estimate the competitiveness of this *randomized online-algorithm.

6.4.1 RONE - AONE with random advice bit

Let's start with AONE from the previous experiment, but guess the adviceBit randomly:

```

1 RONE = class extends AONE
2   oracle: ->
3     [Math.random() < 0.5]

```

If we guess wrong, we might get a lower gain then 0.5 or even 0, if the adviceBit is 1 and we have no item with size > 0.5.

So while we have

```

1 experiments.push
2   name: "RONE - one random bit"
3   description: "Is 4-competitive in expectation"
4   beta: 0.55
5   Algorithm: RONE
6
7 A1 = Greedy
8 A2 = class extends Algorithm
9   reset: ->
10     super
11     @a1 = new A1
12     @doAct "simulateA1"
13
14   decide: (item) ->
15     if @acts "simulateA1"
16       if @a1.handle item
17         no
18       else
19         @doAct "greedy"
20         @decide item
21     else if @acts "greedy"
22       decideGreedy.call @, item
23

```

```

24
25
26
27
28
29
30 RONE2 = class extends Algorithm
31   constructor: ->
32     @a1 = new A1
33     @a2 = new A2
34     super
35   oracle: -> [Math.random() < 0.5]
36   reset: ->
37     super
38     @a1.reset()
39     @a2.reset()
40   knapsack: -> @algorithm().knapsack()
41   handle: (item) ->
42     adviceBit = @readAdviceBit item.index
43     if adviceBit? # existence
44       if adviceBit then @doAct "A1" else @doAct "A2"
45     @algorithm().handle item
46
47   algorithm: ->
48     if @acts "A1" then @a1 else @a2
49
50
51
52
53
54 experiments.push
55   name: "RONE2 - one random bit"
56   description: "Is 2-competitive in expectation"
57   beta: 0.55
58   Algorithm: RONE2
59
60
61
62 createItems = ({beta, maxSize}) ->
63
64   items = []
65   beta ?= 0.5
66   maxSize ?= 1
67   totalSize = 0
68
69   loop
70     randomValue = -> roundValue Math.random()*beta
71     value = randomValue()
72     if totalSize+value < maxSize
73       totalSize += value
74       items.push {value, isPartOfSolution: yes}
75     else
76       # add one that fits exactly
77       items.push
78         value: roundValue maxSize - totalSize
79         isPartOfSolution: yes
80       # add the one that does not fit
81       items.push {value}
82
83
84   break

```



```

85
86
87     items = _.shuffle items
88     for item, index in items
89         item.index = index
90     return items.reverse() # we later pop the elements out (from the
91                             end) because it is faster
92
93 Template.experiments.helpers
94     experiments: -> experiments
95
96
97
98 Template.Experiment.onCreated ->
99
100     @items = []
101
102     @currentItem = new ReactiveVar
103     @numberOfItems = new ReactiveVar
104     @algorithm = new @data.Algorithm
105     @gainHistory =
106         history: []
107         dep: new Tracker.Dependency
108         add: (gainValue) ->
109             if gainValue > 0
110                 @worstGain = Math.min @worstGain ? gainValue,
111                                     gainValue
112                 @bestGain = Math.max @bestGain ? gainValue, gainValue
113                 @history.push gainValue
114                 @dep.changed()
115         size: ->
116             @dep.depend()
117             @history.length
118         worst: ->
119             @dep.depend()
120             @worstGain
121         best: ->
122             @dep.depend()
123             @bestGain
124         competitiveCount: ->
125             @dep.depend()
126             _.countBy @history, (value) ->
127                 if value is 1
128                     "1-competitive"
129                 else if 0.5 <= value < 1
130                     "2-competitive"
131                 else if 0.25 <= value < 0.5
132                     "4-competitive"
133                 else
134                     "non-competitive"
135
136
137         competitivePercentage: (cGroup)->
138             @dep.depend()
139             if @history.length > 0
140                 roundValue 100 * @competitiveCount()[cGroup] /
141                             @history.length
142         avg: ->
143             @dep.depend()

```

```

143         if @history.length > 0
144             roundValue (_,reduce @history, (total, value) -> total
145                 +value)/@history.length
146         reset: ->
147             @history = []
148             @bestGain = null
149             @worstGain = null
150             @dep.changed()
151         resetExperiment = =>
152
153             @items = createItems beta: @data.beta
154             @algorithm.reset?()
155             @algorithm.askOracle? @items
156             @numberOfItems.set @items.length
157             @currentItem.set @items.pop()
158
159         do reset = =>
160             @gainHistory.reset()
161             resetExperiment()
162
163         @ticker = new Ticker
164         reset: =>
165             reset()
166
167         turn: =>
168             # 1. step: fetch new item
169             # 2. step: put it in knapsack
170
171             item = @currentItem.get()
172             if item?
173                 @algorithm.handle item
174                 @currentItem.set @items.pop()
175             else
176                 # no more items
177
178                 @gainHistory.add @algorithm.knapsack().gain()
179                 resetExperiment()
180
181
182         Template.Experiment.helpers
183         adviceBits: -> Template.instance().algorithm.adviceBits.get()
184         act: -> Template.instance().algorithm.act.get()
185         knapsack: -> Template.instance().algorithm.knapsack()
186         ticker: -> Template.instance().ticker
187         currentItem: ->Template.instance().currentItem?.get()
188         gainHistory: -> Template.instance().gainHistory
189         numberOfItems: -> Template.instance().numberOfItems.get()
190         willMatch: ->
191             ctx = Template.instance()
192             ctx.currentItem?.get()?.value + ctx.algorithm.knapsack().gain
193                 () <= ctx.algorithm.knapsack().size
194
195         Template.Knapsack.helpers
196         totalWidth: ->
197             @size * Constants.SCALE + 2
198         items: ->
199             @getItems()
200
201         Template.KnapsackItem.helpers
202         width: ->
203             @value * Constants.SCALE

```

```
202     color: ->
203         hue = @value*360
204         "hsl({hue}, 73%, 69%)"
205
206 Template.TickerGui.helpers
207     counter: -> @ticker.getCounter()
208 Template.TickerGui.events
209     'click .btn-step': -> @ticker.step()
210     'click .btn-play': ->
211         @ticker.setTimeout 100
212         @ticker.play()
213     'click .btn-play-fast': ->
214         @ticker.setTimeout 0
215         @ticker.play()
216
217     'click .btn-stop': -> @ticker.stop()
218     'click .btn-reset': -> @ticker.reset()
```