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# Examine Technology Gaps in Algorithm Structures

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**Yiqiao Yin**  
Ph.D. Student

## Abstract

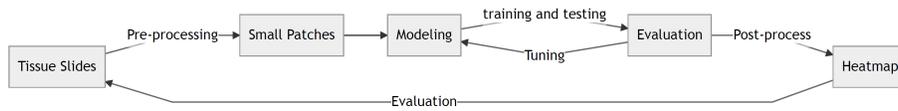
1 This report explores the concepts of novel data ordering techniques.  
2 Particularly, the paper investigates the new alternatives in data struc-  
3 ture, data ordering, data processing, and data base management. The  
4 paper provides some technical discussion and present some sample  
5 code to demonstrate the argument.

## 6 1 Introduction of Algorithm Structure

7 Database management system and data platform play important roles for today's de-  
8 velopment in software engineering and database ordering and management. A famous  
9 example is the Enterprise Resources Planning System (ERP system) which is originally  
10 developed from the Manufacturing Resources Planning (MRP system) (Moon, 2007;  
11 Al-Mashari, 2003; O'Leary, 2004). These generations of development focus on pro-  
12 viding methodologies to all resources in every different component of a company. It is  
13 not surprising that the letter "E" from the name means "enterprise". This is because, in  
14 theory, the platform is designed to serve all levels in organization, financial planning, and  
15 business operations. In practice, ERP systems are capable of integrating inventory data  
16 which is designed to operate on the cloud and also via a variety of different locations.  
17 Just like each business practice has its own unique optimized environment, the ERP  
18 system needs to be able to adapt every building block of an enterprise or a corporation.

19 Another famous example is the Electronic health record or (EHR) system. It is an  
20 systemic collection of patients data designed to store and manage all available health  
21 information in a digital approach (Majeed et al., 2008; Safran and Goldberg, 2000;  
22 Garets and Davis, 2005). These information serve as the basis of health care setting  
23 where physicians and doctors can review information of patients throughout a secured  
24 network program. At enterprise-wide setup, the information is also able to be shared  
25 across different practice therapeutic areas as well as different medical practices. This is  
26 a famous example because data forms are stored completely different in every single  
27 branch of the hospitals and there are insurance companies involved to interfere the  
28 data collection process. The EHR system is capable of storing data and capture the  
29 state of the patient across time. Due to this setup, the tedious label of tracking down  
30 the patients' health data is largely omitted which makes the diagnosis procedure much  
31 more efficient than before. However, the premise is that the EHR system is indeed  
32 capable of the flexibility of editing and modifying patients' health data without leaking  
33 important information to an unrelated third party. This also calls for the needs of point-

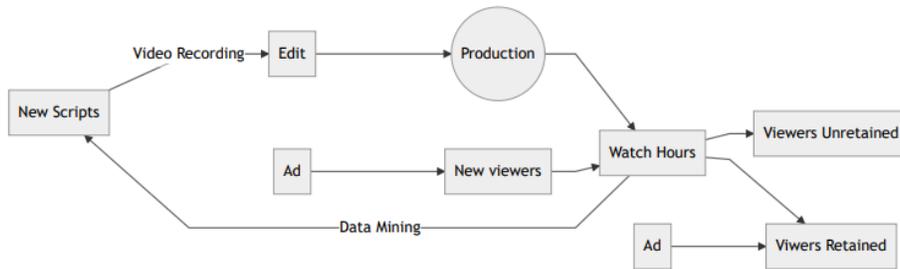
Figure 1: Executive Diagram for Digital Pathology.



34 to-care if high quality of data and healthcare services are provided. In other words, the  
 35 implementation of such EHR system is a key objective for many hospitals and healthcare  
 36 related organizations. An executive diagram of digital pathology is presented in Figure  
 37 1. The tissue samples are collected and exported as images. These images are SVS files  
 38 that are large-scale, i.e. 4-8 GB in sizes. The pre-processing is required to generate  
 39 small patches of pixel values from the original image. The patches are used for training  
 40 set and the models are built. The model has different tuning parameters of which need  
 41 to be tuned throughout a rigorous evaluation procedure. Data structures are converted  
 42 from images to arrays in the modeling process. The output, masks or annotations, is then  
 43 converted back into image format to create heatmap such that images can be generated  
 44 to deploy to machine for doctors to evaluate. Not all image slides are evaluated correctly.  
 45 Hence, this requires more evaluation procedure which can require us to start from the  
 46 pre-preprocessing step.

47 A third example is the Customer Relationship Management (CRM). This type of tech-  
 48 nology is crucial at all operational levels. It is the key component of managing the  
 49 organization’s relationships and interactions between managers and customers. Allow  
 50 me to use my own story as an example. The YouTube channel YinsCapital was founded  
 51 2 years ago. Initially the channel was created for fun and it is designed for leisure  
 52 purpose. However, it has come to the founders’ attention that some videos are doing  
 53 well in terms of watch hours. Hence, this calls an interest to start a data mining project  
 54 on the old videos posted. YouTube provides tabular data that consist of features such  
 55 as "likes", "viewable rates", "cards clicked" and so on. These features have some con-  
 56 tributing power to push for high watch hours which can become Ad sales. Once realized  
 57 the root of the key objectives, a CRM system was immediately developed to ensure  
 58 high customers retention rate and consistent watch hours. A diagram of this executive  
 59 platform is presented in Figure 2. In other words, a CRM platform is required to set up  
 60 for to develop more interesting videos.

Figure 2: Executive Diagram of YouTube CRM.



61 In the beginning stages, the scripts of the new stories are created on the spot without any  
 62 motivation to retain any customers or viewers. This is risky business practice, however,

63 it is the only viable approach without data mining. A series of videos have been posted  
64 and made into the production. Hence, these videos are generating views. The watch  
65 hours can be collected and the attributes of the videos along with watch hours can be  
66 recorded as well. This forms the initial version of the customer database that can be  
67 helpful for R&D team to conduct the experiments using machine learning techniques.  
68 Regression models can be built to predict the watch hours given the attributes of a video.  
69 Inferences can also be made using statistical approach to explain what are the important  
70 features to trigger high watch hours. These results are collected and send to the Director  
71 of film production to push more scripts following the same topic which is supported  
72 by empirical evidence resulting from pattern recognition team using machine learning  
73 algorithms. The key objective is to increase watch hours so that the business can produce  
74 revenue and income. The essence to increase watch hours is to attract new viewers and  
75 also retain old viewers.

76 Now that the business operation and activities are cleared defined, the role of CRM  
77 system needs to follow. The data structure is a key because the video along with related  
78 attributes are essential for data mining. The data from scripts are text format. When the  
79 story goes through recording and editing to production, it will become video format.  
80 The final objective is the watch hours which can simply be a numerical form. Hence, to  
81 keep things simple, denote the script to be  $\mathcal{S}$ , the video to be  $\mathcal{V}$ , and the attributes to be  
82  $\mathcal{X}$ . The watch hours  $Y \in \mathbb{R}$  can be modeled using the following equation

$$Y = f(\mathcal{S}, \mathcal{V}, \mathcal{X}) \quad (1)$$

83 where  $f(\cdot)$  is a function that can be sought using machine learning.

## 84 **2 Motivation to Explore Novel Alternatives for Data structuring** 85 **and Algorithmic Design**

86 Disregard the areas of practice (healthcare, technology, finance, or else), the data structure  
87 and data ordering are crucial components in a development pipeline. As discussed in the  
88 above portion of this report, the ERP system, the EHR system, and the CRM platform  
89 all require different structural design to handle different forms of data. In addition, it is  
90 important to review some famous algorithmic design to understand the motivation of  
91 why it is desirable to come up with novel techniques.

### 92 **2.1 Problem Statement**

93 In algorithmic design and data ordering and structuring, the first key concept is to  
94 understand the key-value pairs. This is referring to the one-one map between the key  
95 and its corresponding contents also commonly known as Hash Table. From practical  
96 perspective, the important property for hash table is that there should not be any collision.  
97 When handling a collision issue, it is either directly resolve via a different style of coding  
98 or the algorithmic can allow multiple items in the same bucket to form the content of  
99 that key.

100 The second types of problem is linked lists. In data structure and algorithmic design,  
101 a linked list is usually a list of nodes. Each node carries attributes such as “value” and  
102 “next” where the “next” stores the address of the next node where “value” stores the  
103 content of what is inside of this node. The problem with linked lists is that they are  
104 tricky to handle and cannot always be easily manipulated. Sometimes the chained lists  
105 or linked lists are even hard coded to avoid code crash. The third type of problem that  
106 can raise potential issues is “heap”. In data structure and algorithmic design, heaps are

107 the type of data that rarely exists and is definitely more esoteric data structure. Usually,  
108 scholars do not do sufficient research in this type of data structures and they can barely  
109 be broadly implemented. However, there is one type of practical area that uses heap  
110 structure is finding the top  $K$  elements in some unsorted array.

## 111 2.2 Potential Impact of the Problem

112 The first problem to address is the collision issue in the algorithmic design. In Figure 1  
113 and Figure 2, both diagrams presented operational and business activities without the  
114 strict premise stating all data structure must be one-one map. In other words, the issue of  
115 collision problem in hash tables exist because not all real world data contain the required  
116 assumption for hash tables to operate properly.

117 For the second problem described above, i.e. the chained lists or the linked lists, it is a  
118 common problem in model building. In both Figure 1 and Figure 2, modeling component  
119 posts great importance in the entire business operation. This is unavoidable and it is a  
120 “must” for the system to learn from the past data. However, a model such as an Artificial  
121 Neural Network (ANN) can sometimes be soft coded using functional API instead of  
122 sequential API. This would be a real world scenario that linked lists play into effect. If  
123 the data structure is not properly managed, the neural network or ANN model could fall  
124 apart. The last problem described above is the “heap” structure. The “heap” structure  
125 is designed to search for the top elements in some unsorted array according to certain  
126 threshold. Most likely, programmers prefer to work with sorted lists. In this case, “heap”  
127 produces the same ideal items without sorting the array. In regards to the consumption  
128 time, it is not as ideal, because it is  $\mathcal{O}(k \log(n))$ . In addition, for unlucky situation, a  
129 complete sort must have  $\mathcal{O}(n \log(n))$  which is not trivial and can be costly to run.

## 130 2.3 Algorithmic Design

131 This section presents some potential algorithmic design addressing each and every  
132 problems stated above.

133 First, it is discussed above that the hash tables have potential collision problems which  
134 is almost always inevitable because the previous section of the table listed out popular  
135 real world scenarios that do not necessarily possess the premises required for hash tables  
136 to exist. One way to work around this concept is the following. Some pseudo code can  
137 be presented below. A hash table can simply be designed using dictionary, i.e. this is the  
138 curly bracket “{}” in python syntax. The for loop dictates the index every step of the  
139 way so that the item can properly be extracted one by one.

```
140 hash_table = {} # this is a dictionary  
141  
142 for i in hash_table:  
143     print(i)
```

145 Each item in the dictionary does not have to have the same length or even the same  
146 object. For example, the same dictionary defined above can be expanded using a list of  
147 integers or a combination of integers and letters. The for loop dictates the output to be  
148 each item from the list.

```
149 hash_table = {  
150     'a': [1,2],  
151     'b': [1,2,3,'c']  
152 } # this is a dictionary  
153 for i in hash_table:  
154
```

```

155 | print(i)
156 |
157 | # output
158 | [1,2] \\  

159 | [1,2,3,'c'] \\  

160 |

```

161 For the second issue regarding linked or chained list, a potential solution with some  
162 basic algebraic operation can be presented below. The example presents a simply defined  
163 node that is written in a class object. The “ThisNode” class object has “value” and  
164 “next” defined in the initiation. Then there is a while loop inside and the while loop  
165 iterative adds value to an output. A function, “list\_to\_linkedlist” is created to execute  
166 an operation using “ThisNode” object. The function checks each item in the entered  
167 number and checks that whether it is equivalent with the “head” or not. It returns the  
168 updated value “head” in the end.

```

169 | class ThisNode(object):
170 |     def __init__(self, value, next=None):
171 |         self.value = value
172 |         self.next = next
173 |
174 |
175 |     def __str__(self):
176 |         curr = self
177 |         output = ''
178 |         while curr:
179 |             output += str(curr.value)
180 |             curr = curr.next
181 |         return output
182 |
183 | def list_to_linkedlist(some_list):
184 |     head = None
185 |     curr = None
186 |     for n in some_list:
187 |         if not head:
188 |             head = ThisNode(n)
189 |             curr = head
190 |         else:
191 |             curr.next = ThisNode(n)
192 |             curr = curr.next
193 |     return head
194 |
195 | n = ThisNode(1, ThisNode(2, ThisNode(3)))
196 | print(n)
197 |
198 | print(list([10, 20, 30]))
199 |

```

200 In machine learning when using ANN, suggested by the previous section, functional API  
201 can sometimes be used to take care of repetitive coding. Some sample code is presented  
202 below. The algorithm starts with “inputs” of which the dimension should be defined. The  
203 hidden layers are tricky to be defined if the user desires to make it soft coded. In other  
204 words, user can enter any list of integers in the beginning and the algorithm automatically  
205 updates the hidden layers of the neural network model.

```

206 | # parameter
207 | params = [128, 64, 32, 10]
208 |
209 |

```

```

210 # initiate the model input
211 inputs = tf.keras.Input(...)
212
213 # the first layer
214 dense = layers.Dense(params[0], activation="relu")
215 x = dense(inputs)
216
217 # hidden layers
218 for i in params[1::]:
219     x = layers.Dense(params[i], activation="relu")(x)
220
221 # the last layer
222 outputs = layers.Dense(params[10])(x)
223
224 # finalized the model
225 model = keras.Model(inputs=inputs, outputs=outputs, name="
226     ↪ mnist_model")

```

228 For “heap” data structure, one approach is to work with queues and it is a good start  
229 because most languages contain built-in support functions. Take python programming  
230 as an example, the heap structure can be imported using “heapq” library. For example,  
231 as the sample code shown below, a list of unsorted numbers can be defined using the  
232 following code. Then the “heapq” library has the “heapify” function that can directly  
233 produce a heap structure.

```

234 import heapq
235
236
237 some_list = [5, 40, 3, 1, 2, 70, 8]
238 heapq.heapify(some_list)
239 print(some_list)
240 print(heapq.heappop(some_list))
241 print(heapq.heappop(some_list))
242 print(heapq.heappop(some_list))
243 print(heapq.heappop(some_list))
244 print(heapq.heappop(some_list))
245 print(heapq.heappop(some_list))
246 print(heapq.heappop(some_list))

```

248 To sort a heap structure, the code below is a good example, along with the hand-written  
249 implementation of a heap. This is considered one of the most complicated algorithm,  
250 but it is fairly straightforward upon reading it.

```

251 def heapify(arr, n, i):
252     largest = i # Initialize largest as root
253     left = 2 * i + 1 # left = 2*i + 1
254     right = 2 * i + 2 # right = 2*i + 2
255
256
257     # See if left child of root exists and is
258     # greater than root
259     if left < n and arr[largest] < arr[left]:
260         largest = left
261
262     # See if right child of root exists and is
263     # greater than root
264     if right < n and arr[largest] < arr[right]:

```

```

265     largest = right
266
267     # Change root, if needed
268     if largest != i:
269         arr[i],arr[largest] = arr[largest],arr[i] # swap
270
271         # Heapify the root.
272         heapify(arr, n, largest)
273
274 # The main function to sort an array of given size
275 def heapSort(arr):
276     n = len(arr)
277
278     # Build a maxheap.
279     for i in range(n, -1, -1):
280         heapify(arr, n, i)
281
282     # One by one extract elements
283     for i in range(n-1, 0, -1):
284         arr[i], arr[0] = arr[0], arr[i] # swap
285         heapify(arr, i, 0)
286
287 # Driver code to test above
288 arr = [ 12, 11, 13, 5, 6, 7]
289 heapSort(arr)
290 n = len(arr)
291 for i in range(n):
292     print ("%d" %arr[i], end=' ')
293 # 5 6 7 11 12 13
294

```

### 295 **3 Conclusion**

296 This report examined famous algorithm structures to identify some of the famous research  
297 problems in data structure and algorithmic design. Famous structures include (1) hash  
298 tables, (2) linked lists, and (3) heaps. The paper started with the introductory section and  
299 explained with live examples in the ERP, MRP, and CRM system the issues and social  
300 impact of data structures. Then the report stated the problems from technical prospective.  
301 The report listed out the potential impact of these data structures based on the technical  
302 component and produced some code samples to illustrate the root of the problems.

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