



VRIJE
UNIVERSITEIT
BRUSSEL



ML Algorithm Interview Preparation 2020

INTERVIEW PREPARATION

LiangLiang ZHENG

Academic Year: 2019-2020

Promotor:

Advisor:

Set a faculty using \faculty{\textit{Engineering Sciences}}

Contents

1 Machine Learning & Statistics	1
1.1 Linear Regression	1
1.1.1 Probabilistic interpretation	2
1.1.2 K-fold CV & Stratified K-fold CV	3
1.2 Accuracy, Precision, Recall	3
1.3 Statistical Tests Other	3
1.3.1 Null Hypothesis	3
1.3.2 Correlational Test	4
1.3.3 Comparison of Means test	4
1.3.4 Non-parametric Test	4
1.4 ROC	4
1.5 Cross-Validation	4
2 Natural Language Processing	5
2.1 Language Model	5
2.1.1 Unigram, Bi-Gram, etc	5
2.1.2 What are word embeddings?	5
2.1.3 Context Window	6
2.1.4 Two Types of Embedding Models	7
2.1.4.1 CBOW	7
2.1.4.2 Skip-Gram	7
2.2 TF-IDF	7
3 Deep Learning	9
3.1 GNN	9
4 Math	11
5 Algorithm	13
5.1 Time Complexity and Space Complexity	13
5.2 General Algorithmic Knowledge	13
5.2.1 Base Conversion	13
5.2.2 Concatenate bits	13
5.2.3 Max Counter	13
5.2.4 alphabet changing	13
5.2.5 check valid parentheses	13
5.2.6 Time Complexity and Space Complexity	14
5.2.7 Bit operations, Mod and Pow	14
5.2.8 count frequency	14
5.2.9 python: indices even and odd sub list	14

5.2.10 python: even and odd	14
5.3 Basic Data Structure	15
5.3.1 Coding Template	15
5.3.1.1 Binary Search	15
5.3.1.2 Tree	15
5.3.2 Array	16
5.3.3 Linked List	16
5.3.3.1 commonly used tricks	16
5.3.3.2 Reverse Linked List	17
5.3.3.3 Fast-slow pointer	21
5.3.3.4 Advanced problems, not frequently asked questions	22
5.3.4 Stack	23
5.3.5 Queue	23
5.3.6 Deque	23
5.3.7 Tree	23
5.3.7.1 summary	23
5.3.7.2 Traversal	24
5.3.7.3 Binary Tree Pruning	32
5.3.7.4 DFS related questions	33
5.3.7.5 Construct Tree	35
5.3.7.6 LCA	36
5.3.7.7 Tree serialization and de-serialization	37
5.3.7.8 Binary Search Tree	37
5.4 Advanced Data Structure	40
5.4.1 Priority Queue	40
5.4.2 Trie	40
5.4.3 Segment Tree	40
5.5 Search	40
5.5.1 Backtrack	40
5.5.1.1 combination	41
5.5.1.2 subset	42
5.5.1.3 permutation	43
5.5.2 BFS	44
5.5.3 DFS	50
5.5.3.1 Sudoku, N-Queue	51
5.5.3.2 DFS+BFS	54
5.6 Binary Search	56
5.7 Dynamic Programming	64
5.7.1 Simple Example: Fibonacci	65
5.7.2 Coin Change	65
5.7.3 Longest Increasing Subsequence (LIS) $O(n^2)$	66
5.7.4 Maximum Subarray	68
5.7.5 01 Knapsack problem	71

5.7.6	Subset Knapsack problem	71
5.7.7	Complete Knapsack Problem	73
5.7.8	Edit Distance	74
5.7.9	Longest Common Subsequence(LCS)	75
5.7.10	Longest Palindromic Subsequence	75
5.7.11	Super Egg Drop	78
5.7.12	Burst Balloons	80
5.7.13	Stock Buying Selling Problem	82
5.7.14	House Robber	87
5.7.15	RegExp Match	89
5.7.16	Knuth-Morris-Pratt (KMP)	91
5.8	Two Pointers	91
5.9	HashTable	91
6	Tensorflow & PyTorch	93
7	Spark	95
7.1	Rationale of using Spark	95
7.2	Core Concept and How Spark Work	96
7.2.1	Spark Component	97
7.2.2	Spark Architecture	98
7.3	Spark Programming	99
7.3.1	Difference between SparkSession and SparkContext	99
8	Docker & Kubernetes	101
8.1	Useful commands	102
8.2	Docker compose	103
8.3	Docker workflow & Develop with docker	104
8.4	Dockerfile	105
8.5	Private Docker Repository	106
8.6	Deploy and persistence	106

Chapter 1

Machine Learning & Statistics

1.1 Linear Regression

The difference between **classification** and **regression** problem is depending on the target variable, if the target variable is continuous, we call it regression, otherwise, it is a classification problem.

To perform a supervised learning, we need to formulate an hypotheses h which approximate our real \mathbf{y} . A simple linear regression of the h is formulated as followed:

$$h_{\theta}(x) = \theta_0 + \theta_1 x_1 + \theta_2 x_2 \quad (1.1.1)$$

Here, the θ_i 's are the parameters (also called weights) parameterizing the space of linear functions mapping from \mathcal{X} to \mathcal{Y} . When there is no risk of confusion, we will drop the θ subscript in $h_{\theta}(x)$, and write it more simply as $h(x)$. To simplify our notation, we also introduce the convention of letting $x_0 = 1$ (this is the intercept term).

The cost function is defined as followed, this is call least square cost function, the choice of using least square comes from the fact that it'll be easier to calculate the derivatives of the cost function.

$$J(\theta) = \frac{1}{2} \sum_{i=1}^n \left(h_{\theta}(x^{(i)}) - y^{(i)} \right)^2 \quad (1.1.2)$$

What we need to computer to do is to minimize $J(\theta)$ as small as possible, we use **Gradient Descent**, what gradient descent dose is starts with an 'initial guess', and and that repeatedly changes to make $J(\theta)$ smaller. α is the learning rate, the reason that is minus instead of add the partial derivative of (θ) is that, you could think of (θ) as the graph of x^2 , you start with a random x , every step you're going downward with the direction of the slope of x^2 .

$$\theta_j := \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta) \quad (1.1.3)$$

The break down of the partial derivative of $J(\theta)$ is as followed:

$$\begin{aligned}
\frac{\partial}{\partial \theta_j} J(\theta) &= \frac{\partial}{\partial \theta_j} \frac{1}{2} (h_{\theta}(x) - y)^2 \\
&= 2 \cdot \frac{1}{2} (h_{\theta}(x) - y) \cdot \frac{\partial}{\partial \theta_j} (h_{\theta}(x) - y) \\
&= (h_{\theta}(x) - y) \cdot \frac{\partial}{\partial \theta_j} \left(\sum_{i=0}^d \theta_i x_i - y \right) \\
&= (h_{\theta}(x) - y) x_j
\end{aligned} \tag{1.1.4}$$

As you can see from above, when calculating the partial derivatives of $J(\theta)$ wrt θ_j , the least square actually brings some convenience, in the end what we get, intuitively speaking, is the error between hypothesis h and y with respect to the current x we're looking at. Instead of looking at each training set each time of the iteration, **Batch Gradient Descent** looks at every example in the entire training set on every step.

1.1.1 Probabilistic interpretation

In the above of examples we have known that the form of the cost function (least square) could come from that fact it'll be easier for us to calculate the partial derivatives in the gradient descent phase. Another reason could be derived from the probabilistic perspective.

The relationship of the target hypothesis and real value y could be formed as

$$y^{(i)} = \theta^T x^{(i)} + \epsilon^{(i)}$$

where $\epsilon^{(i)}$ is an error term that captures either unmodeled effects, or random noise. Let us further assume that the $\epsilon^{(i)}$ are distributed IID (independently and identically distributed) according to a Gaussian distribution (also called a Normal distribution) with mean zero and some variance σ^2 . We can write this assumption as " $\epsilon^{(i)} \sim \mathcal{N}(0, \sigma^2)$ ". I.e., the density of $\epsilon^{(i)}$ is given by

$$p(\epsilon^{(i)}) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(\epsilon^{(i)})^2}{2\sigma^2}\right).$$

This implies that

$$p(y^{(i)} | x^{(i)}; \theta) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(y^{(i)} - \theta^T x^{(i)})^2}{2\sigma^2}\right)$$

When we wish to explicitly view this as a function of θ , we will instead call it the likelihood function:

$$L(\theta) = L(\theta; X, \vec{y}) = p(\vec{y} | X; \theta) \tag{1.1.5}$$

The MLE (Maximization of Likelihood Estimation) is a method we used to do the parameter estimation, here we have the parameter theta to be estimated, in order to

maximize the likelihood, normally we took the log of this likelihood function (because likelihood function involves tons of probability products, using log form to transform it to summation form), so we have:

$$\begin{aligned}
 L(\theta) &= \prod_{i=1}^n p(y^{(i)} | x^{(i)}; \theta) \\
 &= \prod_{i=1}^n \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(y^{(i)} - \theta^T x^{(i)})^2}{2\sigma^2}\right) \\
 \ell(\theta) &= \log L(\theta) \\
 &= \log \prod_{i=1}^n \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(y^{(i)} - \theta^T x^{(i)})^2}{2\sigma^2}\right) \\
 &= \sum_{i=1}^n \log \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(y^{(i)} - \theta^T x^{(i)})^2}{2\sigma^2}\right) \\
 &= n \log \frac{1}{\sqrt{2\pi}\sigma} - \frac{1}{\sigma^2} \cdot \frac{1}{2} \sum_{i=1}^n (y^{(i)} - \theta^T x^{(i)})^2.
 \end{aligned} \tag{1.1.6}$$

Hence, maximizing $\ell(\theta)$ gives the same answer as minimizing

$$\frac{1}{2} \sum_{i=1}^n (y^{(i)} - \theta^T x^{(i)})^2$$

1.1.2 K-fold CV & Stratified K-fold CV

1.2 Accuracy, Precision, Recall

1.3 Statistical Tests Other

1.3.1 Null Hypothesis

Given the test scores of two random samples, one of men and one of women, does one group differ from the other? A possible null hypothesis is that the mean male score is the same as the mean female score: $H_0 : \mu_1 = \mu_2$ where

$$\begin{aligned}
 H_0 &= \text{the null hypothesis,} \\
 \mu_1 &= \text{the mean of population 1, and} \\
 \mu_2 &= \text{the mean of population 2.}
 \end{aligned}$$

A stronger null hypothesis is that the two samples are drawn from the same population, such that the variances and shapes of the distributions are also equal.

1.3.2 Correlational Test

- Pearson correlation: Tests for the strength of the association between two **continuous variables**.
- Spearman correlation: Tests for the strength of the association between two **ordinal variables** (does not rely on the assumption of normal distributed data).
- Chi-square: Tests for the strength of the association between two **categorical variables**.

1.3.3 Comparison of Means test

T-Test Assumptions

- The first assumption made regarding t-tests concerns the scale of measurement. The assumption for a t-test is that the scale of measurement applied to the data collected follows a continuous or ordinal scale, such as the scores for an IQ test.
- The second assumption made is that of a simple random sample, that the data is collected from a representative, randomly selected portion of the total population.
- The third assumption is the data, when plotted, results in a normal distribution, bell-shaped distribution curve.
- The final assumption is the homogeneity of variance. Homogeneous, or equal, variance exists when the standard deviations of samples are approximately equal.

Paired T-test: Tests for difference between two related variables. Independent T-test: Tests for difference between two independent variables. ANOVA: Tests the difference between group means after any other variance in the outcome variable is accounted for.

1.3.4 Non-parametric Test

Wilcoxon rank-sum test: Tests for difference between two independent variables - takes into account magnitude and direction of difference. **Wilcoxon sign-rank test:** Tests for difference between two related variables - takes into account magnitude and direction of difference. **Sign test:** Tests if two related variables are different – ignores magnitude of change, only takes into account direction.

1.4 ROC

1.5 Cross-Validation

Chapter 2

Natural Language Processing

2.1 Language Model

2.1.1 Unigram, Bi-Gram, etc

We need a model that will assign a probability to a sequence of tokens. Let us start with an example: "The cat jumped over the puddle." A good language model will give this sentence a high probability because this is a completely valid sentence, syntactically and semantically. Similarly, the sentence "stock boil fish is toy" should have a very low probability because it makes no sense. Mathematically, we can call this probability on any given sequence of n words:

$$P(w_1, w_2, \dots, w_n)$$

We can take the unary language model approach and break apart this probability by assuming the word occurrences are completely independent:

$$P(w_1, w_2, \dots, w_n) = \prod_{i=1}^n P(w_i)$$

However, we know this is a bit ludicrous because we know the next word is highly contingent upon the previous sequence of words. And the silly sentence example might actually score highly. So perhaps we let the probability of the sequence depend on the pairwise probability of a word in the sequence and the word next to it. We call this the bigram model and represent it as:

$$P(w_1, w_2, \dots, w_n) = \prod_{i=2}^n P(w_i | w_{i-1}) \quad (2.1.1)$$

2.1.2 What are word embeddings?

In their most basic form, word embeddings are a technique for identifying similarities between words in a corpus by using some type of model to predict the co-occurrence of

words within a small chunk of text. By examining the adjacency of words in this space, word embedding models can complete analogies such as “Man is to woman as king is to queen.”

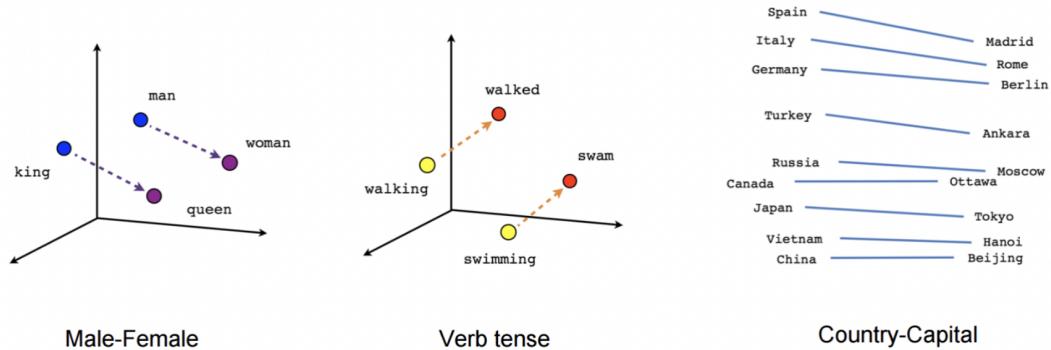


Figure 2.1: word embedding

2.1.3 Context Window

Word embeddings are created by identifying the words that occur within something called a “Context Window.” The Figure below illustrates context windows of varied length for a single sentence. The context window is defined by a string of words before and after a focal or “center” word that will be used to train a word embedding model. Each center word and context words can be represented as a vector of numbers that describe the presence or absence of unique words within a dataset, which is perhaps why word embedding models are often described as “word vector” models, or “word2vec” models.

- : Center Word
- : Context Word

- c=0 The cute **cat** jumps over the lazy dog.
- c=1 The **cute** **cat** jumps over the lazy dog.
- c=2 The **cute** **cat** jumps over the **lazy** dog.

Figure 2.2: context window

2.1.4 Two Types of Embedding Models

Word embeddings are usually performed in one of two ways: “Continuous Bag of Words” (CBOW) or a “Skip-Gram Model.” The figure below illustrates the differences between the two models. The CBOW model reads in the context window words and tries to **predict the most likely center word**. The Skip-Gram Model **predicts the context words given the center word**. The examples above were created using the Skip-Gram model, which is perhaps most useful for people who want to identify patterns within texts to represent them in multidimensional space, whereas the CBOW model is more useful in practical applications such as predictive web search.

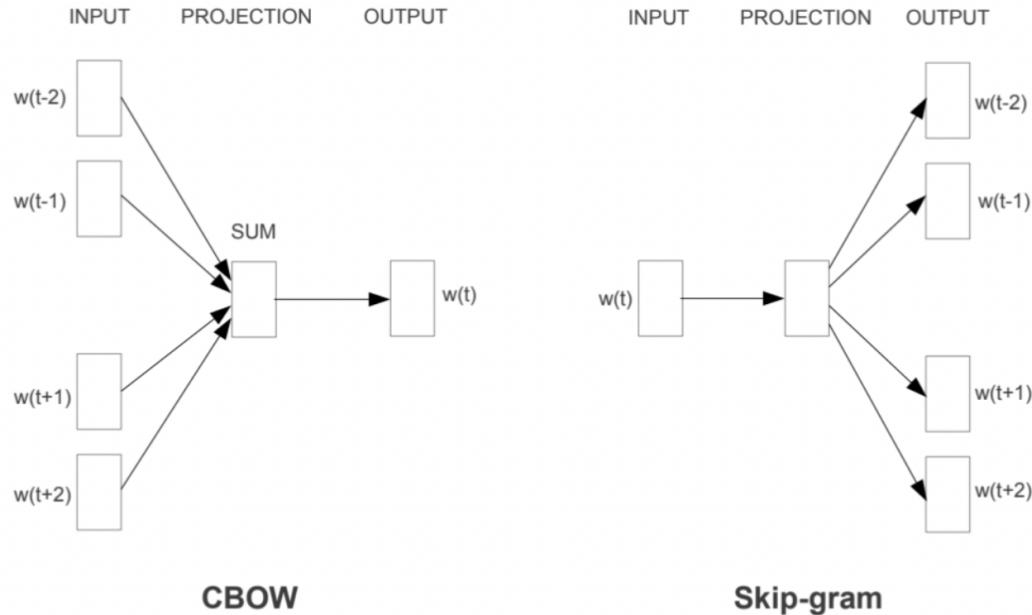


Figure 2.3: CBOW v.s. Skip Gram

2.1.4.1 CBOW

2.1.4.2 Skip-Gram

2.2 TF-IDF

Chapter 3

Deep Learning

3.1 GNN

Representation Learning: Automatically extract or learn the feature, the idea is to map d-dimensional embeddings such that similar nodes in the network are embedded close together.

Chapter 4

Math

Chapter 5

Algorithm

5.1 Time Complexity and Space Complexity

5.2 General Algorithmic Knowledge

5.2.1 Base Conversion

- Binary of a number: `bin(a)`
- Binary to `int(res, 2)`

5.2.2 Concatenate bits

The verticle line | in python is "bitwise or" method, it'll return 1 if one of it is 1.

`s << i.bit_length()|i`

5.2.3 Max Counter

```
l = len(nums)
a = collections.Counter(nums)
return a.most_common(1)[0][0]
```

5.2.4 alphabet changing

To know if a character is a alphabet, `'a'.isalpha()`

5.2.5 check valid parentheses

Check if a sequence of parentheses are valid, count the left paren first and if a right paren occurs and the left paren is 0, which means before this right paren there must be no left paren that could be corresponded to it. The code below is the code, which will be used in #301.

```

def valid(s):
    l = 0
    r = 0
    for i in s:
        if i == '(':
            l += 1
        if i == ')' and l == 0:
            r += 1
        elif i == ')':
            l -= 1
    return l == 0 and r == 0

```

5.2.6 Time Complexity and Space Complexity

5.2.7 Bit operations, Mod and Pow

Alphabet case conversion XOR operation with ' ', for example, converting 'd' to 'D' or do it reversely, simply do `chr(ord('d')^ord(''))`

In #952, mod is always used in circle like number like circular list. A circle like number 0-9 where $0 - 1 = 9$ and $9 + 1 = 0$, the easiest way to know what number to mod could produce 0, in this case only 10 could do that.

5.2.8 count frequency

```

self.count = collections.Counter()
self.inorder(root)
freq = max(self.count.values())
res = []
for item, c in self.count.items():
    if c == freq:
        res.append(item)
return res

```

5.2.9 python: indices even and odd sub list

```

even = nums[::2]
odd = nums[1::2]

```

5.2.10 python: even and odd

```

even = a % 2 == 0 or even = a & 1 == 0
odd = a % 2 == 1 or odd = a & 1 == 1

```

5.3 Basic Data Structure

5.3.1 Coding Template

5.3.1.1 Binary Search

```
def left_bound(nums, target):
    l = 0
    r = len(nums)
    while l < r:
        mid = l + (r - 1) // 2
        if nums[mid] >= target:
            r = mid
        elif nums[mid] < target:
            l = mid + 1

    return l

def right_bound(nums, target):
    l = 0
    r = len(nums)
    while l < r:
        mid = l + (r - 1) // 2
        if nums[mid] <= target:
            l = mid + 1
        else:
            r = mid

    return l - 1
```

5.3.1.2 Tree

inorder traversal: recursion
 inorder traversal: iterative
 pre-order and postorder: recursion
 pre-order and postorder: iterative
 level-order iterative

dfs build graph: template for turning a tree to a non-directed graph

```
graph = collections.defaultdict(list)
def dfs(parent, child):
    if parent and child:
        graph[parent.val].append(child.val)
```

```

graph[child.val].append(parent.val)

if child.left: dfs(child, child.left)
if child.right: dfs(child, child.right)
dfs(None, root)

```

5.3.2 Array

5.3.3 Linked List

5.3.3.1 commonly used tricks

- dummy as previous node of head (#2, #445, #24)
- if not head or not head.next (#206, #24)
- fast-slow pointers (#141, #142, #876)

Example: In **Leetcode#2**, after transferring list to number using the function **l2n**, the result has been transferred to a string, in this way every element inside can be gotten simply by using index number, and in the end the list can be built in a iterative way. A dummy is initialized as head and will be used as return pointer because in iteration head is already changed. For loop is backward, From len - 2 (len - 1 should be assigned to head before) to the end 0.

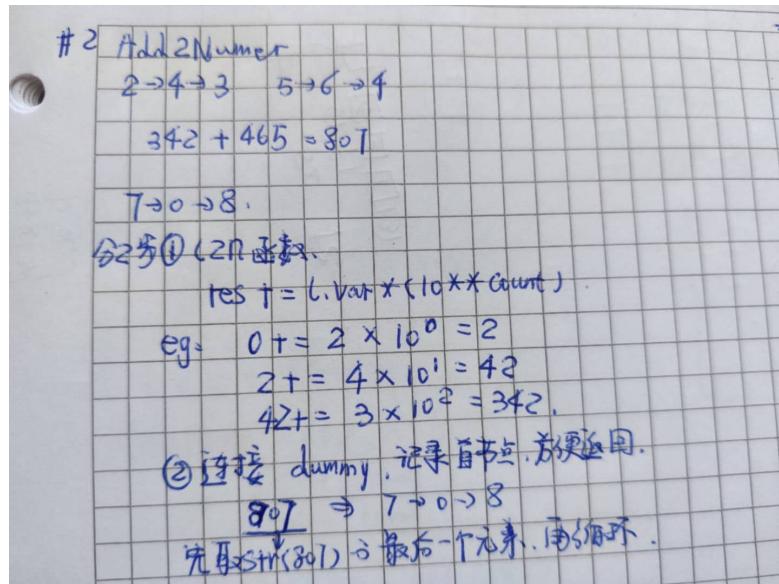


Figure 5.1: two sum

```

class Solution(object):
    def addTwoNumbers(self, l1, l2):

```

```

"""
:type l1: ListNode
:type l2: ListNode
:rtype: ListNode
"""

def l2n(l):
    res = 0
    count = 0
    while l:
        res += l.val * (10 ** count)
        l = l.next
        count = count + 1

    return res

str_add = str(l2n(l1) + l2n(l2))
head = ListNode(str_add[len(str_add) - 1])

dummy = head

for i in range(len(str_add) - 2, -1, -1):
    node = ListNode(str_add[i])
    head.next = node
    head = head.next

return dummy

```

Leetcode#445, add two number II, the only difference is that the function **l2n** is difference. $res = res * 10 + l.val$, and in for loop in index i is forward, from 1 (0 should be assigned to head before) to the end len - 1.

5.3.3.2 Reverse Linked List

Leetcode#206 is to reverse linked list, there are iterative way and recursive way to reverse the linked list. The first one below is **iterative** way, the important message is that whenever you want to assign a pointer.next to other node, don't forget to put pointer.next to a new variable in this way we won't lose it. In this problem, it's important to know whether head is None or not, we also need to know head.next for the reason that if this list only has 1 node.

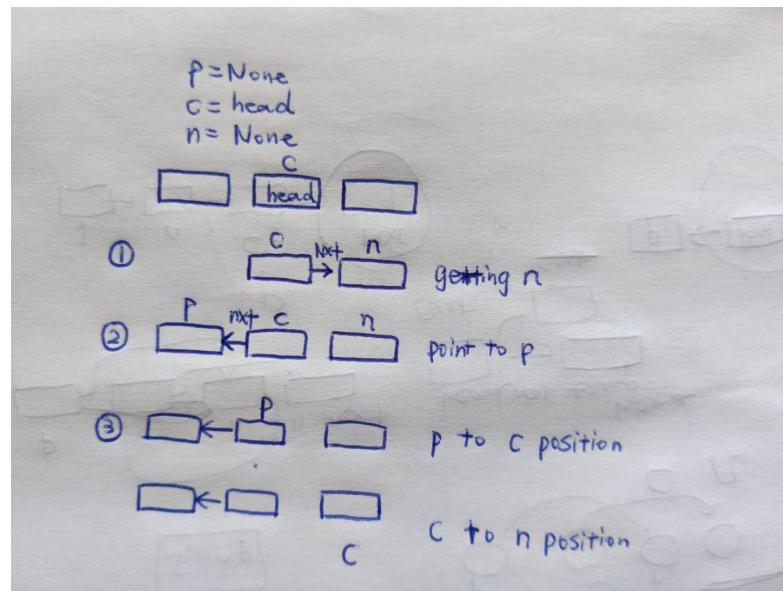


Figure 5.2: reverse list in iterative way

```

class Solution(object):
    def reverseList(self, head):
        if not head or not head.next:
            return head

        p = None
        c = head
        n = None

        while c:
            n = c.next
            c.next = p
            p = c
            c = n

        return p
    
```

For **recursive** way, `reverseListBlog` explained very clearly, the Figure shown below also illustrated the process of reverse a linked list in a recursive way.

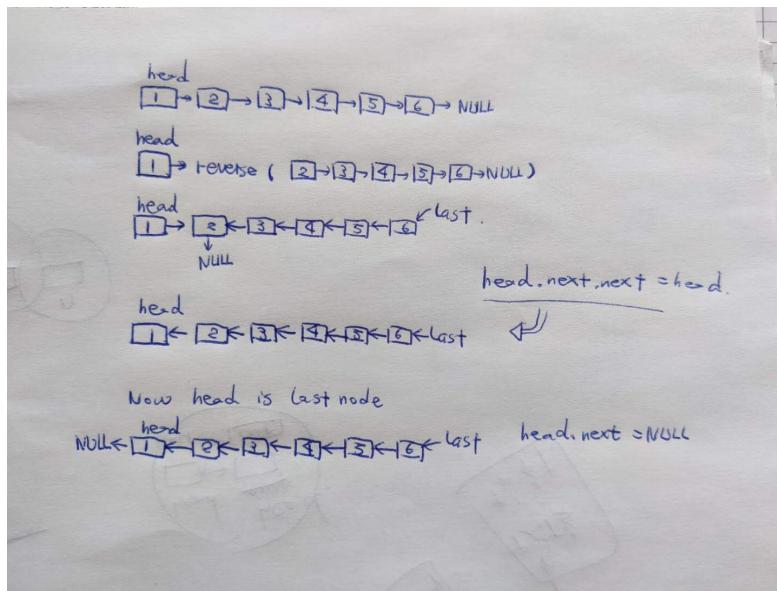


Figure 5.3: reverse list in recursive way

```
def reverseList(self, head):
    """
    :type head: ListNode
    :rtype: ListNode
    """
    if not head or not head.next:
        return head

    last = self.reverseList(head.next)
    head.next.next = head
    head.next = None

    return last
```

If we want to reverse the first N node, it is similar to the implementation of the recursive one, only need to know the successor, the successor is the one right after node $N + 1$, for example, 1-2-3-4-5-6-None, will return as 3-2-1 4-5-6, but in this case 1.next is not None anymore, it should be 4, so in recursive we need to know the node and assign that in a variable **successor**. The next is the Figure of Reverse N node, and code below the figure is the **Leetcode#92**

```
class Solution(object):
    successor = None
    def reverseBetween(self, head, m, n):
        def reverseN(head, n):
```

```

global successor
if n == 1:
    successor = head.next
    return head

last = reverseN(head.next, n - 1)
head.next.next = head
head.next = successor

return last

if m == 1:
    # same as reverse first N node
    return reverseN(head, n)

# forward as to the base case m = 1
head.next = self.reverseBetween(head.next, m - 1, n - 1)
return head

```

For Leetcode#24, swap nodes between nodes should could be solved with iterative way using 2 pointers n1 and n2, the graph and code is shown below.

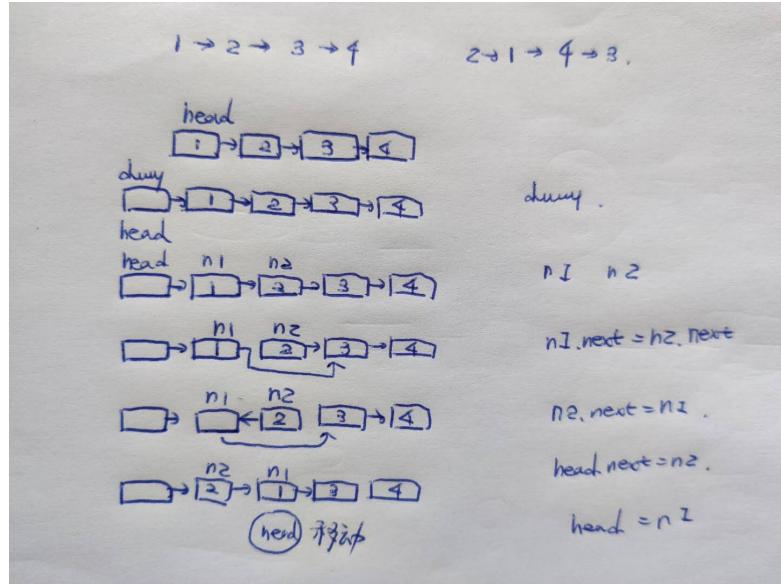


Figure 5.4: swap 2 nodes

```

def swapPairs(self, head):
    if not head or not head.next:
        return head

```

```

dummy = ListNode(-1)
dummy.next = head
head = dummy

while head.next and head.next.next:
    n1,n2 = head.next, head.next.next
    n1.next = n2.next
    n2.next = n1
    head.next = n2
    head = n1

return dummy.next

```

5.3.3.3 Fast-slow pointer

Leetcode#141 Fast-slow pointer always used in check if a linked list has cycle, the fast pointer will go 2 step at a time while the slow pointer will go 1 step at a time. The initial step is to assign head to both of them, in while loop, the condition is while(fast.next), this is because if reach the second last node, fast.next is the last one, for the iteration after this will end.

```

def hasCycle(self, head):
    slow = head
    fast = head

    while fast and fast.next:
        fast = fast.next.next
        slow = slow.next

        if fast == slow:
            return True

    return False

```

Leetcode#142, the only difference is to know the starting node of the cycle, the following graph shows after k step, slow and fast meet each other in the Meeting Point, the starting point should before m step of Meeting Point, after putting slow back to head, let slow and fast go 1 step at a time, after k - m step, they will meet at the Starting point. (Remeber there are 2 ways to go out of the while loop, the first one is break, the second one is normal running, then in that case there is no Cycle inside, we shoud return None instead).

```

def detectCycle(self, head):
    slow = head

```

```

fast = head

while fast and fast.next:
    fast = fast.next.next
    slow = slow.next

    if slow == fast:
        break
else:#No Cycle (out of loop not because of break)
    return None

slow = head
while slow != fast:
    slow = slow.next
    fast = fast.next

return fast

```

Leetcode#876 is an easy one, same as above, when fast reaches the end, slow will be the middle node.

5.3.3.4 Advanced problems, not frequently asked questions

Leetcode#25 could be solved using reverseBetween, 2 passes in solving this, first is to count the length of the list, and the second is to iteratively call reverseBetween with corresponding index m and n. If the k is 2, then it is similar to **Leetcode#24**

```

def reverseKGroup(self, head, k):
    if (head is None or k == 1):
        return head

    # First, get the length of the list
    length = self.getCount(head)

    # Second pass, swap in groups.
    newH = head
    for i in range(1, length+1, k):
        if i+k-1 <= length:
            newH = self.reverseBetween(newH, i, i+k-1)

    return newH

def getCount(self, node):
    n = node

```

```

count = 0
while n:
    count+=1
    n = n.next
return count

def reverseN(self,head, n):
    global suc
    if n == 1:
        suc = head.next
    return head

    last = self.reverseN(head.next, n-1)
    head.next.next = head
    head.next = suc
    return last

def reverseBetween(self,head, m, n):
    if m == 1:
        return self.reverseN(head, n)

    head.next = self.reverseBetween(head.next, m - 1, n - 1)

    return head

```

5.3.4 Stack

5.3.5 Queue

5.3.6 Deque

5.3.7 Tree

5.3.7.1 summary

- stack for doing the iterative form of traversal, the pattern of thinking should be reverse. Specifically, for inorder traversal, left-root-right, find left most first, but on the way to find left, stacking root also, since after getting left most, pop out order is left -> root. so stack.append(root) should be applied before finding left most node. **Memorization Trick: Y true not stack**
- For pre-order, root-left-right, unlike pop should be right before stack.append, since the stacking is reverse, stacking right first and then left branch. For postorder, the code is similar, left-right-root, we do it in a reverse way root-right-left, then

the question becomes similar to pre-order, after getting res list, reverse it and then return.**Memorization Trick: Y stack**

- For Level-order(BFS), its better to use collections.deque(), which will save more time, the list.pop(0) have complexity of O(n).
- construct trees
- Getting the depth of a tree, normally using recursion, get left and right and consider returning a new value, which normally will be the max(left, right)+1 or other kinds of form. If there are other restriction in the tree depth, like finding the smallest, or finding compare the tree depth, then we to do something after getting and l and r.
- There are several problem set use both children but only return one child when doing the recursion, for example, #124, The definition of the path is **a path is defined as any sequence of nodes from some starting node to any node in the tree along the parent-child connections**, means it could go through only left or right, but in the description Example, it shows that add both children only if the starting node of this path is their direct parent. If this is not the case, it can only return one of its children. So in the coding part. the trick is **res can append both children but only return one**. for #534 and #687 are also the same.
- PathSUM problem
- Go through the root or dose not go through the root? From root to the leaf, should give condition on the

Different kind of tree

- Binary Search Tree: the inorder traversal is sorted. Those 2 algorithms are bonded tightly
-

5.3.7.2 Traversal

Leetcode #94 is a classic **inorder** traversal of tree question, the following is the template of recursively do the inorder traversal.

```
def inorderTraversal(self, root):
    # recursive way
    res = []

    def inorder(root):
        # base case
        if not root:
```

```

        return root

    inorder(root.left)
    res.append(root.val)
    inorder(root.right)

inorder(root)
return res

```

The recursive way is shown below, the intuition behind the algorithm is that we want to get the leftmost child first in the res list, using stack here to push the previous nodes inside until we find the first leftmost child (means it has no left child anymore). **Leetcode #173** is also using iterative way to inorderly traverse a Binary Search Tree.

```

def inorderTraversal(self, root):
    # iterative way
    res = [] # the result order
    stack = [] #

    while True:
        while root:
            # find the left most
            stack.append(root)
            root = root.left
        # if there was no element inside stack, similar to if not root in recursive version
        if not stack:
            return res

        # found the left most child
        left = stack.pop()
        res.append(left.val)
        root = left.right

```

Leetcode #589, **Leetcode#114** is the **preorder** and **Leetcode #590** is the **postorder**.

```

def preorder(self, root):
    # recursive way
    res = []

    def preorder1(root):
        if not root:
            return

        res.append(root.val)

```

```

    for c in root.children:
        preorder1(c)

    preorder1(root)
    return res

```

The following is the answer for #114, since the output should be a linked list tree, then the res.append should append input TreeNode instead of root.val, and assign the next element inside the list to the right node and the left node to None.

```

def flatten(self, root):
    # preorder recursive way
    res = []

    def preorder(root):
        if not root:
            return root

        res.append(root)
        preorder(root.left)
        preorder(root.right)

    preorder(root)

    for i in range(len(res) - 1):
        res[i].left = None
        res[i].right = res[i+1]

    return res

```

And the iterative way is as followed, after res.append, we need to reverse the way of the output, specifically it means that stack the one that needs to output later first, in this case it's the last child of the children, so we could simply use node.children[::-1] in this case, and use extend to get rid of the square braces problems.

```

def preorder(self, root):
    # iterative way
    res = []

    if root == None:
        return res

    stack = []
    stack.append(root)

```

```

while stack:
    node = stack.pop()
    if not node:
        continue
    # reverse the way to output, so stack in right first
    res.append(node.val)
    # stack.append(node.right)
    # stack.append(node.left)
    stack.extend(node.children[::-1])
return res

```

The following is the graph shows how it works in general

The serialization of the preorder is also a problem set, for example in **leetcode #331**, the form of the preorder is always follows number # #.

```

def isValidSerialization(self, preorder):

    #the leaves will always follow the form of number # #
    # only need to replace the last 3 element to # and iterate the list

    stack = collections.deque()

    pre = preorder.split(",")

    for i in pre:
        print(i)
        stack.append(i)

    while len(stack) >= 3 and stack[-1] == stack[-2] == '#' and stack[-3] != '#':
        stack.pop()
        stack.pop()
        stack.pop()
        stack.append('#')

    return len(stack) == 1 and stack.pop() == '#'

```

For postorder, the recursive way is the same as inorder and preorder except the order of append root.val, the following is the solution for **Leetcode #590**. The form is similar to the iterative form of preorder, but stacking node.left and then node.right, and reverse all the res list.

```

def postorder(self, root):
    # iterative way

```

```

res = []
if root == None:
    return res
stack = []
stack.append(root)
while stack:
    node = stack.pop()
    # if not node:
    #     continue
    res.append(node.val)
    stack.extend(node.children[:])

res.reverse()
return res

```

recursive way for implementing the traversal is actually important in this section, since they are basic framework (or say template) to solve other kinds of problem.

```

def recursive(root):
    if not root:
        return
    # do what you want to do with root
    recursive(root.left)
    recursive(root.right)

```

In **Leetcode #100**, in order to know whether 2 tree are the same, in the recursion, first take out their root and compare. The recursion is also something will be covered in DFS, where the fist line in the recursion are always those end case. In sameTree, the end condition is when 2 nodes are None, if not, then check if one of them is empty. if the structure are the same, the final condition is whether the values are the same. This ensemble the war defense, soldiers set up layer to defend enermies.

```

def isSameTree(self, p, q):
    # both of them are None, only one case with output True
    if p is None and q is None:
        return True
    #one of it is None
    if p is None or q is None:
        return False
    if p.val != q.val:
        return False

    return self.isSameTree(p.left, q.left) and self.isSameTree(p.right, q.right)

```

For **Leetcode #101**, same as isSameTree, only difference is change recursion to opposite node. self.isSameTree(p.left, q.right) and self.isSameTree(p.right, q.left)

For **Leetcode #104** finding the maximum depth of the tree, simply three lines could solve it. Same pattern as the traversal recursion. When root is None, return 0, otherwise recursively call the max (left and right) + 1.

```
def maxDepth(self, root):
    if not root:
        return 0
    return max(self.maxDepth(root.left), self.maxDepth(root.right)) + 1
```

For minimum (#111) depth of the tree, after recursively calling the function, there are 2 situations, which are drawn below, when it comes to leave, 2 nodes like the graph should be counted as 2, so $l + r + 1 = 2$, while the other condition is similar to max.

```
def minDepth(self, root):
    if not root:
        return 0

    l = self.minDepth(root.left)
    r = self.minDepth(root.right)
    # be careful! if there are only 2 nodes
    if l == 0 or r == 0:
        return l + r + 1
    else:
        return min(l, r) + 1
```

Same for #110, get height using maxDepth function as in #104 and add one condition $\text{abs}(l - r) \neq 1$ into the recursion call.

In the next subsection, we're going to discuss Level order traversal, this is also a sub pre-requisite for Search part BFS. The following code is the template for level order traversal(BFS).

```
def level_order_tree(root, result):
    if not root:
        return
    # using collections.deque()
    # avoid using list.pop(0) since the complexity is O(n)
    queue = collections.deque()
    queue.append(root)
    while queue:
        node = queue.popleft()
        # do somethings
        result.append(node.val)
        if node.left:
```

```

        queue.append(node.left)
    if node.right:
        queue.append(node.right)
    return result

```

In most of the leetcode problem set, we need to output according to layers, so inside the while, we need to have a for loop and a newQueue which could do it in different layer . #102 and #107 are using the template below.

```

def level_order_tree(root):
    if not root:
        return
    q = [root]
    while q:
        # normally res will be here
        new_q = []
        for node in q:
            # do somethings with this layer nodes eg: 103, set flag to append in zigzag
            # classical condiction
            if node.left:
                new_q.append(node.left)
            if node.right:
                new_q.append(node.right)
        # replace as the new queue
        q = new_q
    # return things you want
    return xxx

```

#987 is a good problem set, it combines the level-order problem with coordinate questions. The first need to watch out is the form inside queue, should be (node, x, y), the order of the value and node is not matter, while inside res, we should put x and y in front since in the end of the while, res should be sorted in this way we'll could put those with the same coordinate in one cell from left to right. up to down, so y should be incremented instead of decremented as mentioned in the problem description.

```

def verticalTraversal(self, root):
    res = []
    if not root:
        return res

    q = []
    #initial coordinates
    q.append((root, 0, 0))

    while q:

```

```

        node, x, y = q.pop()
        res.append((x, y, node.val))
        # no need to do it in differen layer
        # so no for loop here
        if node.left:
            q.append((node.left, x-1,y+1))
        if node.right:
            q.append((node.right, x+1,y+1))

    res.sort()
    print(res)
    # put root inside the res1 first
    res1 = [[res[0][2]]]
    # res2 = []
    for i in range(1, len(res)):
        if res[i][0] == res[i - 1][0]:
            res1[-1].append(res[i][2])
        else:
            res1.append([res[i][2]])

    return res1

```

#116 and #117

```

def connect(self, root):
    """
    :type root: Node
    :rtype: Node
    """
    # Level-order way with iterative
    res = []
    if not root:
        return None

    q = collections.deque()
    q.append(root)

    while q:
        newQ = collections.deque()
        for i in range(len(q)):
            n = q.popleft()
            res.append(n.val)
            if n.left:
                newQ.append(n.left)

```

```

        if n.right:
            newQ.append(n.right)
            res.append('#')

q = newQ

a = Node(res[0])
p = a
for i in range(1, len(res)-1):
    if res != '#':
        b = Node(res[i])
    else:
        b = None
    a.next = b
    a = b
return p

```

5.3.7.3 Binary Tree Pruning

Leetcode #814 is still a postorder, could use DFS(recursive way) to do it, since we need to know it from the leaves and then bottom up pruning the tree, the return value should be a node, the one to get the return value (condition) should be root.left and root.right. Trim it in the last step after all the conditions.

```

def pruneTree(self, root):
    if not root:
        return
    root.left = self.pruneTree(root.left)
    root.right = self.pruneTree(root.right)
    # only return node with 1 inside or not leaves
    return root if root.val == 1 or root.left or root.right else None

```

For **Leetcode #669**, trim it when compare to L and R and get left and right in the last condition.

```

def trimBST(self, root, L, R):

    if not root:
        return

    if root.val < L:
        return self.trimBST(root.right, L, R)

    elif root.val > R:
        return self.trimBST(root.left, L, R)

```

```

    else:
        root.left = self.trimBST(root.left, L, R)
        root.right = self.trimBST(root.right, L, R)
        return root

```

5.3.7.4 DFS related questions

A lot of problem sets could be solved in a recursive way. The followings are some problems that is typical recursion-based question. It is noticeable to know the nature of recursion, recursion is a structure of tree, every branch (or same iteration) will go through the deepest leave and if found no node after, then backtrack and return. So the first couple lines of DFS are always condition, the first line will also be the last condition, for example in tree the first condition is always to know whether this node is empty or not to see if this recursion has reached the end of the tree, then next couple of conditions will be based on specific question set, the last is to recursively called the function

Leetcode #112 and **Leetcode #113** are among them, the difference of them is that in 112 there is only one answer while in 113 there are multiple answers, then we need a path + [root.left.val] inside the recursion, since we call the root.left.val which means we assume root.left is indeed exist, we can't wait until the next iteration to check whether root.left is existed, then in this round, it's also necessary to check if root.left is existed. The 2 templates below could be used in any path sum related questions.

```

def hasPathSum(self, root, s):

    if not root:
        return False

    if root.left == None and root.right == None and s == root.val:
        return True
    # either branch is okay, so OR is used here
    return self.hasPathSum(root.left, s - root.val) or self.hasPathSum(root.right, s - root.val)

def pathSum(self, root, s):
    res = []
    if not root:
        return res

    def helper(root, s, path):
        if not root:
            return
        if root.left == None and root.right == None and s == root.val:

```

```

        res.append(path)
        return

    if root.left:
        helper(root.left, s - root.val, path + [root.left.val])
    if root.right:
        helper(root.right, s - root.val, path + [root.right.val])

helper(root, s, [root.val])

return res

```

Leetcode#437 is also among them. But 437 has several answer and which may not from root to leave, which also need to recurse starting from every node below. then there are 2 recursion running below. First is the helper which did as the same as what we need above, and called pathSum twice and recursively call root.left and root.right inorder to start with a new child node to search.

```

def pathSum(self, root, s):
    if not root:
        return 0
    return self.helper(root, s) + self.pathSum(root.left, s) + self.pathSum(root.right, s)

def helper(self, root, s):
    if not root:
        return 0
    res = 0
    if s == root.val:
        res += 1

    res += self.helper(root.left, s - root.val)
    res += self.helper(root.right, s - root.val)
    return res

```

The next part is MAX PathSUM related questions: Use both children, but return one, these kind of question is a little bit tricky, but it can also be solved using DFS way. For **Leetcode#124**, the graph has been shown below, we need to find the maximum path sum, the end condition is not go through the leaves, so we need no condition to stop until it reaches the end of the tree, but the tricky part is the result layer could be $root.val+l+r$ while in recursion, we can only choose one of the path(biggest path) to search. And in case of negative number, l and r should be compared to 0.

```

def maxPathSum(self, root):
    self.res = float('-inf')

```

```

def maxPath(root):
    if not root:
        return 0

    l = max(0, maxPath(root.left))
    r = max(0, maxPath(root.right))
    # final answer should add l + r + root.val
    self.res = max(l + r + root.val, self.res)
    # while this return will go the next layer, only need one path
    return max(l, r) + root.val

maxPath(root)
return self.res

```

leetcode #235 and **leetcode #236** is also part of these problems, like what we did in maxPath, divide(recursively called and get l and r first) and conquer (and then do something with the result l and r).

```

def lowestCommonAncestor(self, root, p, q):
    if any((not root, root == p, root == q)):
        return root

    l = self.lowestCommonAncestor(root.left, p, q)
    r = self.lowestCommonAncestor(root.right, p, q)

    #means there are l and r with same root
    if l and r:
        return root
    # if not, return one of the branch which has and recursively call again
    else:
        return l if l else r

```

508

5.3.7.5 Construct Tree

The approach to construct tree is very simple, given the rule(BST, then mid is root)(max tree, build according to max in every recursive array), and slice the array for left branch and right branch inside the two recursive expression shown below.

```

def buildTree(self, preorder, inorder):
    if not preorder or not inorder:
        return None

```

```

# first element in preorder is root, found the index in preorder and separate
root = TreeNode(preorder[0])
index = inorder.index(preorder[0])
# length of left branch in preorder is 1(root) to index + 1,
root.left = self.buildTree(preorder[1:index+1], inorder[:index])
root.right = self.buildTree(preorder[index+1:], inorder[index+1:])
return root

```

5.3.7.6 LCA

These kinds of problem sets are a little bit difficult. 236 and 235 is the LCA problem, the recursion thinking pattern is to get l and r first, if found it, then could be return. since the return value is the definition of the function. But this process is after the recursion, the condition that could find the result is before the recursion, when the root is None, means not root, then return root, if the root is equal to p or q(remember this is after the recursion, root.left and root.right has been passed). if both returns a valid node which means p, q are in different subtrees, then root will be their LCA. if only one valid node returns, which means p, q are in the same subtree, return that valid node as their LCA.

```

def lowestCommonAncestor(self, root, p, q):
    if not root or q == root or p == root:
        return root

    l = self.lowestCommonAncestor(root.left, p, q)
    r = self.lowestCommonAncestor(root.right, p, q)

    if l and r:
        return root
    else:
        return l if l else r

```

Iterative Way:

```

def lowestCommonAncestor(self, root, p, q):
    pathP, pathQ = self.findpath(root, p), self.findpath(root, q)
    if pathP and pathQ:
        length = min(len(pathQ), len(pathP))
        ans = None
        for i in range(length):
            if pathP[i] != pathQ[i]:
                break
            ans = pathP[i]
        return ans
    def findpath(self, root, key):
        if not root:

```

```

        return
stack = []
lastVisit = None
while stack or root:
    if root:
        stack.append(root)
        root = root.left
    else:
        tmp = stack[-1]
        if tmp.right and tmp.right != lastVisit:
            root = tmp.right
        else:
            if tmp.val == key.val:
                return stack
            else:
                lastVisit = stack.pop()
                root = None

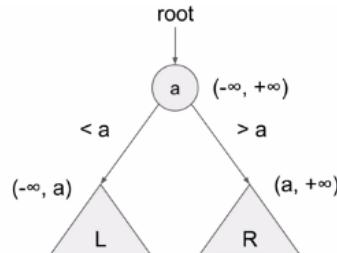
```

5.3.7.7 Tree serialization and de-serialization

#652 #297

5.3.7.8 Binary Search Tree

#98 To validate a binary search tree, normally BST-related questions could be solved using recursion, so this problem could be solved in 2 ways. The first is give limit condition, condition is shown below.



Property of a binary search tree

- $\text{root.val} > \text{all vals in L}$
- $\text{root.val} < \text{all vals in R}$
- Inorder traversal, vals are sorted

Figure 5.5: Validate Binary Search Tree

```

def isValidBST(self, root):
    return self.isvalid(root, float('inf'), float('-inf'))

```

```

def isvalid(self, root, max, min):
    if not root:
        return True

    if root.val >= max or root.val <= min:
        return False

    return self.isvalid(root.left, root.val, min) and \
           self.isvalid(root.right, max, root.val)

```

The second way is to use inorder, the inorder traversal of BST is sorted, so just to know in the recursion if prev is greater or equal to root, then return False, it is also noticeable that the assignment of prev should be after the first recursion.

```

class Solution(object):
    def isValidBST(self, root):
        self.prev = None
        def inorder(root):
            if not root:
                return True
            if not inorder(root.left):
                return False
            if self.prev and self.prev.val >= root.val:
                return False
            self.prev = root
            return inorder(root.right)
        return inorder(root)

```

#701 combining the concept of binary search and build tree, very elegant solution.

```

def insertIntoBST(self, root, val):
    if not root:
        return TreeNode(val)
    if val > root.val:
        root.right = self.insertIntoBST(root.right, val)
    if val < root.val:
        root.left = self.insertIntoBST(root.left, val)
    return root

```

501 is a very good problem, to find the highest frequency element inside the Binary search tree, the first solution is to use collections.Counter() to do that, and there are possibilities that not only 1 highest frequency element, that is why the return should be a list.

```

def findMode(self, root):
    if not root:
        return []
    self.count = collections.Counter()
    self.inorder(root)
    freq = max(self.count.values())
    res = []
    for item, c in self.count.items():
        if c == freq:

```

```

        res.append(item)
    return res

def inorder(self, root):
    if not root: return
    self.inorder(root.left)
    self.count[root.val] += 1
    self.inorder(root.right)

#450,
• The deleted node has no left tree, return right tree
• The deleted node has no right tree, return left tree
• Has both left and right tree, to way to do it
    – Search the min in right tree, swap the nodes and recursively delete the key
    again
    – Search the max in the left tree, swap the nodes and recursively delete the key
    again

def deleteNode(self, root, key):
    if not root: return root
    if root.val == key:
        # if root is a leave
        if not root.right:
            return root.left
        else: # both left and right exist
            # set the min of right tree to be the new node
            right = root.right
            while right.left:
                right = right.left
            right.val, root.val = root.val, right.val
            root.left = self.deleteNode(root.left, key)
            root.right = self.deleteNode(root.right, key)

    return root

```

5.4 Advanced Data Structure

5.4.1 Priority Queue

5.4.2 Trie

5.4.3 Segment Tree

5.5 Search

5.5.1 Backtrack

Summarize from labladong:

- subset: start
- combination: start
- permutation used
- bfs:
- dfs:

Backtracking is the traversal of decision tree in essence. There are 3 things:

- Path: decisions have been made
- Choice list: the choices you can choose now
- End condition: reach to the end of the decision tree

General framework of the backtracking is to do **recursion in for loop, do a choice before the recursion and undo the choice after the recursion**

```
res = []
def backtrack(path,choice_list):
    if end_condition:
        result.append(path)
        return

    for c in choice_list:
        do the choice (select the choice)
        backtrack(path, choice_list)
        undo the choice c (remove the choice from choice_list)
```

5.5.1.1 combination

Input 2 number n and k, output the all combinations of k number in [1..n]. This is also a backtracking problem, but k limit the tree depth, and n limit the width of the tree. Which means the algorithm is similar to subset problem, the only difference is that **k limit the depth, so the end condition should take k into account and end the recursion.** Other than this, combination and subset problem should include **start** index to indicate that the element used before can not be used. There are 2 cases.

- `backtrack(path + [something], i)`: means if combining [1,2,3,4], i = 1 then element 2 can be used in the next round but 1 cannot be used (different from permutation)
- `backtrack(path + [something], i + 1)`: means if combining [1,2,3,4], i = 1 then element 2 can not be used in the next round but 1 cannot be used, it could only start from element 3(index = 2)

Figure 5.6 shown below shows the tree like search process, in some cases, the combination will have constraints like there is **duplicate** inside the list choices but the output cannot have duplicate result, this Figure below(lowest tree) shows it will have same result when it contains duplicate, then an expression `if i > start and c[i] == c[i - 1] continue` should be used, and the choice list should be **sorted** beforehand. Also path + [i] is a technique here work as DFS, the traditional way is to `path.append()` and `path.pop()`. #17 #39 #40 are the combination problem, the template below is the solution of #40 Combination Sum II.

```
def combinationSum2(self, c, t):
    res = []
    if c is None or t == 0:
        return res

    def backtrack(path, target, start):
        if target == 0:
            res.append(path)
            return

        for i in range(start, len(c)):
            if c[i] > target:
                continue
            if i > start and c[i] == c[i - 1]:
                continue
            backtrack(path + [c[i]], target - c[i], i + 1)

    c.sort()
    backtrack([], t, 0)
    return res
```

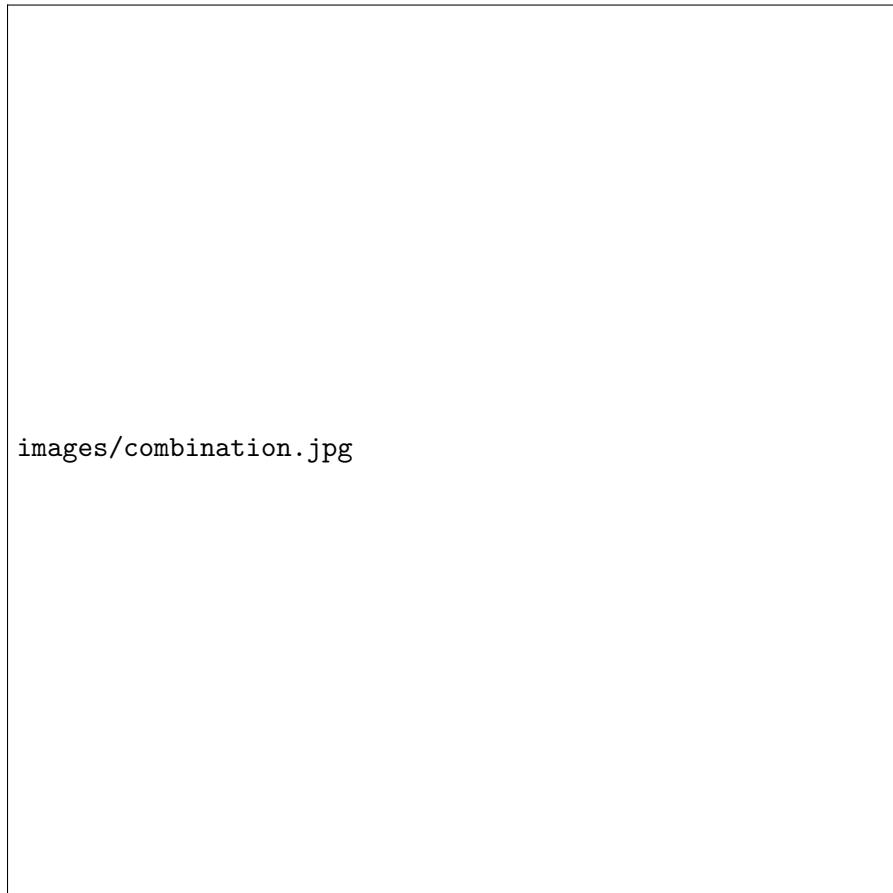


Figure 5.6: Combination

5.5.1.2 subset

concept of these questions: given an array, return its all subsets. eg: `nums = [1,2,3]`, the subsets are `[[],[1],[2],[3],[1,3],[2,3],[1,2],[1,2,3]]`, the order doesn't matter. The question is similar to combination, the only difference is that it doesn't need to have any specific rule to append the path into res.

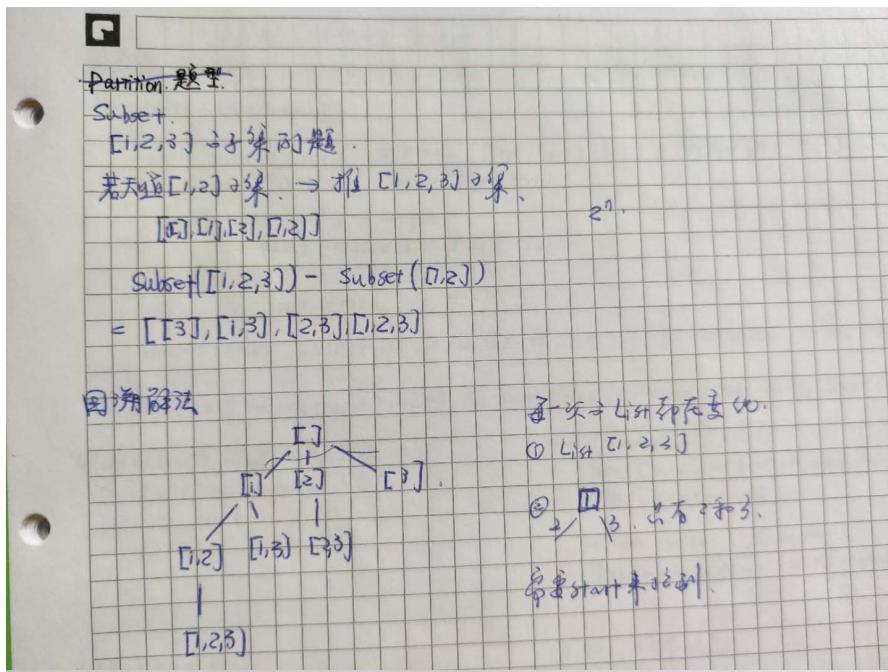


Figure 5.7: subset

5.5.1.3 permutation

The only difference with subset and combination is that the number before could be used, so be careful when doing this, we only need a 'used' array to denote whether the element have been used in this recursion. The following is the solution for #47

An excellent explanation of whether to use 'used[i]' or 'not used[i]' when doing pruning could be found in <https://www.cxyxiaowu.com/2946.html>.

```
def permuteUnique(self, nums):
    res = []
    if nums == None:
        return res
    used = [0] * len(nums)
    def backtrack(path):
        if len(path) == len(nums):
            res.append(path)
            return
        for i in range(len(nums)):
            if i > 0 and nums[i] == nums[i - 1] and not used[i - 1]:
                continue
            if used[i]:
                continue
            used[i] = 1
            backtrack(path + [nums[i]])
            used[i] = 0
```

```

used[i] = 1
backtrack(path + [nums[i]])
used[i] = 0

nums.sort()
backtrack([])
return res

```

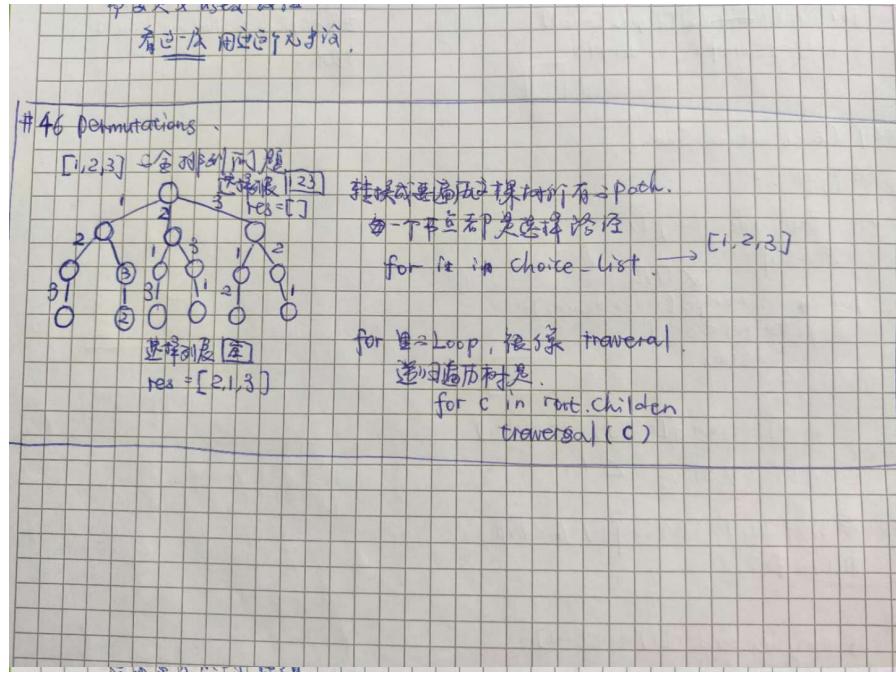


Figure 5.8: Permutation

5.5.2 BFS

The template of BFS is similar to level order traversal of a binary tree, using queue to append root and while loop the root, since we need to get the node level by level, so the size of nodes inside queue should be gained in order to for loop all the nodes inside queue. BFS is a typical **Start to Target** problem, and always is to find the **Shortest Path from Start to the Target**. Recall in level order traversal there is no end condition in for loop, while in BFS, if we found the **Target**, then we should return the step. A typical return value is the value(num of levels to get to the target). Other than that, since this is not a tree anymore, the search direction is not stricted in downward(left node or right node), it could search all of the adjacent nodes, so a visited set should be initialized to ensure the node has not been visted yet. BFS is usually used for solving the least step to solve some search based problem.

The following is the template for BFS:

```
1 // 计算从起点 start 到终点 target 的最近距离
2 int BFS(Node start, Node target) {
3     Queue<Node> q; // 核心数据结构
4     Set<Node> visited; // 避免走回头路
5
6     q.offer(start); // 将起点加入队列
7     visited.add(start);
8     int step = 0; // 记录扩散的步数
9
10    while (q not empty) {
11        int sz = q.size();
12        /* 将当前队列中的所有节点向四周扩散 */
13        for (int i = 0; i < sz; i++) {
14            Node cur = q.poll();
15            /* 划重点：这里判断是否到达终点 */
16            if (cur is target)
17                return step;
18            /* 将 cur 的相邻节点加入队列 */
19            for (Node x : cur.adj())
20                if (x not in visited) {
21                    q.offer(x);
22                    visited.add(x);
23                }
24        }
25        /* 划重点：更新步数在这里 */
26        step++;
27    }
28 }
```

Figure 5.9: BFS

126 word ladder is a typical BFS problem. But in typical BFS problem, normally inside the for loop, we will see if the node is in visited, if it is in visied, then stop this loop with **continue**, while in this case we want to find the node which is in the visited, so those are not in the visited should be ignored. So strictly speaking, this **set** should be declared as wordListSet instead of visited.

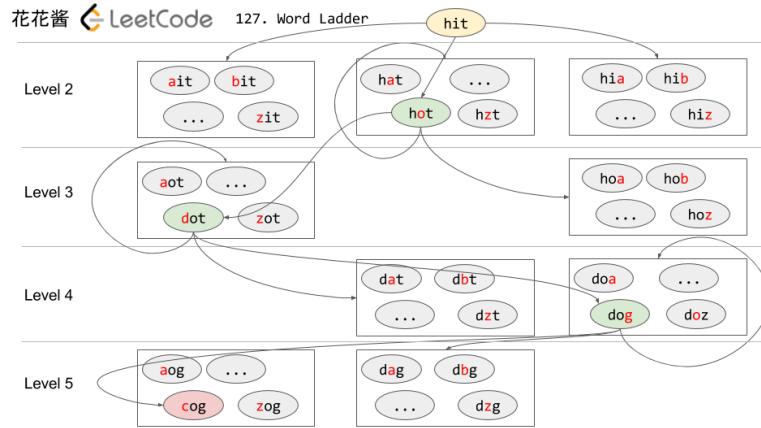


Figure 5.10: wordLadder

```

def ladderLength(self, beginWord, endWord, wordList):
    visted = set(wordList)
    q = collections.deque()
    q.append(beginWord)
    if endWord not in visted: return 0
    step = 0

    while q:
        qSize = len(q)
        step += 1
        for k in range(qSize):
            node = q.popleft()
            for i in range(len(node)):
                for j in "abcdefghijklmnopqrstuvwxyz":
                    word = node[:i] + j + node[i+1:]
                    if endWord == word and word in visted:
                        return step+1

                    if word not in visted:
                        continue

                    if word in visted and word != node:
                        visted.remove(word)
                        q.append(word)
    return 0

```

#752 Open the lock, the first code "0000" should also be put into the deadends set to avoid go back to the origin lock set. simple, but the point of the circle number is important to be remembered. 0 to 9 and 9 to 0, which could be done by mod 10.

```

def openLock(self, deadends, target):
    if "" == target or "0000" in deadends or target in deadends:
        return -1

    q = collections.deque()
    q.append("0000")

    visited = set(deadends)
    visited.add("0000")

    step = 0
    while q:
        step += 1
        qSize = len(q)

        for i in range(qSize):
            node = q.popleft()
            for j in range(len(node)):
                for k in (1, -1):
                    num = (int(node[j]) + k) % 10
                    new = node[:j] + str(num) + node[j+1:]

                    if new == target:
                        return step

                    if new in visited:
                        continue

                    q.append(new)
                    visited.add(new)

    return -1

```

#542 01 Matrix, get every step into the dist function. the interesting thing here is when search in 4 different directions, we could use dirs = [0,-1,0,1,0] to reduce the amount of code we need to write, simply loop in range(4) and add the first to x and add the second to y. This solution below is the one that will not exceed the time limit

The idea of this question is first enqueue all of the positions that are 0, the rest of them are initialized as float('inf'), then use BFS with 4 different directions to find the distance difference that is bigger than 1, then update the dist + 1.

```

def updateMatrix(self, matrix):
    n = len(matrix)
    m = len(matrix[0])

```

```

q = collections.deque()
dist = [[float("inf") for i in range(m)] for j in range(n)]

for i in range(n):
    for j in range(m):
        if matrix[i][j] == 0:
            dist[i][j] = 0
            q.append((i, j))

dirs = [0, -1, 0, 1, 0]
while q:
    x, y = q.popleft()

    for i in range(4):
        a = x + dirs[i]
        b = y + dirs[i+1]
        if 0 <= a < n and 0 <= b < m and dist[a][b] - dist[x][y] > 1:
            dist[a][b] = dist[x][y] + 1
            q.append((a, b))

return dist

```

#675 Cut Off Trees for Golf Event.

```

def cutOffTree(self, f):
    if f == None:
        return -1
    h = []
    # sort the height
    m = len(f)
    n = len(f[0])
    for i in range(m):
        for j in range(n):
            if f[i][j] != 0:
                h.append((f[i][j], i, j))

def bfs(f, sx, sy, tx, ty):
    q = collections.deque()
    q.append((sx, sy))
    visited = [[0 for i in range(n)] for j in range(m)]

    step = 0
    dirs = [0, -1, 0, 1, 0]

```

```

while q:
    qSize = len(q)

    for i in range(qSize):
        (x, y) = q.popleft()
        if x == tx and y == ty:
            return step
        for j in range(4):
            nx = x + dirs[j]
            ny = y + dirs[j + 1]

            # Out of bound or unwalkable
            if (nx < 0 or nx == m or ny < 0 or ny == n or not f[nx][ny] or visited[nx][ny] == 1):
                continue
            q.append((nx, ny))
            visited[nx][ny] = 1
    step += 1

return float('inf')

# sort the height
h.sort()
sx = 0
sy = 0
ts = 0

# loop the height an call BFS
for i in range(len(h)):
    tx = h[i][1]
    ty = h[i][2]

    s = bfs(f, sx, sy, tx, ty)
    if s == float('inf'):
        return -1

    ts += s
    sx = tx
    sy = ty

return ts

```

5.5.3 DFS

DFS is backtracking problem in essence, for #301, this is a hard question, the template is the same as the backtrack problem, the difference is that we need to know how many l and r should be deleted(kinda like target in this case), for every element in list_choice, just simply remove the element at this level and continue the recursion. curr = s[:i] + s[i+1:] will do this action. While the way to verify if a sequence of parentheses are valid is shown in the section 5.2.5.

```

def removeInvalidParentheses(self, s):
    res = []
    if s == "":
        return [""]
    l = 0
    r = 0
    for i in s:
        if i == '(':
            l += 1
        if i == ')' and l == 0:
            r += 1
        elif i == ')':
            l -= 1

    #2. if a string is valid
    def valid(s):
        l = 0
        r = 0
        for i in s:
            if i == '(':
                l += 1
            if i == ')' and l == 0:
                r += 1
            elif i == ')':
                l -= 1
        return l == 0 and r == 0

    def backtrack(s, l, r, start):
        if l == 0 and r == 0 and valid(s):
            res.append(s)
            return

        for i in range(start, len(s)):
            if i > start and s[i] == s[i - 1]:
                continue

```

```

# if s[i] is parenthesis
if s[i] == '(' or s[i] == ')':
    curr = s[:i] + s[i+1:]
# start from right parenthesis
if r > 0:
    # start is i because i is removed, the i+1 now becomes
    # i, so start again with i instead of i + 1
    backtrack(curr, l, r-1, i)
elif l > 0:
    backtrack(curr, l-1, r, i)
backtrack(s, l, r, 0)
return res

```

For #22, same backtracking problem, but the key thing is to understand for loop in backtracking, in this case, there is no need to use for loop, for loop only used in the situation that there are several list choices, in this case, the choices are left or right parentheses, but the actions are taken in different condition, which means it has all choices made already.

```

def generateParenthesis(self, n):
    res = []
    if n == 0:
        return [""]

def backtrack(path, l, r):
    if l > r:
        return

    if l == 0 and r == 0:
        res.append(path)
        return
    if l > 0:
        backtrack(path + '(', l-1, r)
    if r > 0:
        backtrack(path + ')', l, r-1)
backtrack("", n, n)
return res

```

5.5.3.1 Sudoku, N-Queue

#37 Sudoku, this is similar to permutation, this is a brute force DFS, the used array have 3, cols, rows and boxes, the only thing need to be remember is the when found 1 solution, it could return True already, so it won't search other results. And the index of box is also important in this case $x*3 + y$.

```

def solveSudoku(self, board):
    if not board:
        return
    # initialize the board

    rows = [[0 for i in range(10)] for j in range(9)]
    cols = [[0 for i in range(10)] for i in range(9)]
    box = [[0 for i in range(10)] for j in range(9)]

    for i in range(9):
        for j in range(9):
            if board[i][j] != '.':
                a = int(board[i][j])
                rows[i][a] = 1
                cols[j][a] = 1
                box_x = i // 3
                box_y = j // 3
                box[box_x*3 + box_y][a] = 1

    def backtrack(x, y):
        if x == 9:
            return True

        ny = (y + 1) % 9
        if ny == 0:
            nx = x + 1
        else:
            nx = x

        if board[x][y] != '.':
            return backtrack(nx, ny)

        # backtracking
        for i in range(1, 10):
            box_x = x // 3
            box_y = y // 3
            box_index = box_x * 3 + box_y
            if not rows[x][i] and not cols[y][i] and not box[box_index][i]:
                # take actions
                rows[x][i] = 1
                cols[y][i] = 1
                box[box_index][i] = 1

```

```

    board[x][y] = str(i)

    if backtrack(nx, ny):
        return True

    board[x][y] = '.'
    rows[x][i] = 0
    cols[y][i] = 0
    box[box_index][i] = 0
return False

backtrack(0, 0)

```

#51 N-Queue, the ‘used’ array also have 3, rows, diag1 and diag2, in this case, the bracktrack parameter x means cols(rows is also fine). For Sudoku, the choice list have 1 to 9, so for in range(1, 10), but in N-Queue, the choice now becomes cols, so only for in range(n). and we can get corresponding index of those ‘used’ array but only one index ‘i’, so rows, diags and diag2 is only a one dimentional array.

```

def solveNQueens(self, n):
    res = []
    if n == 0:
        return res

    rows = [0] * n
    diag1 = [0] * (2*n - 1)
    diag2 = [0] * (2*n - 1)

#initialize the board
board = [['.' for _ in range(n)] for _ in range(n)]

def backtrack(x):
    if x == n:
        res.append(["".join(i) for i in board])
        return

    for i in range(n):
        if not rows[i] and not diag1[x+i] and not diag2[x - i + n - 1]:
            rows[i] = 1
            diag1[x+i] = 1
            diag2[x - i + n - 1] = 1
            board[x][i] = 'Q'

            backtrack(x+1)

```

```

rows[i] = 0
diag1[x+i] = 0
diag2[x - i + n - 1] = 0
board[x][i] = '.'

backtrack(0)
return res

# Word Search, same backtracking, but be careful when used this element in the
# board and set it to nil and set it back after the backtrack. Otherwise this element might
# be used again in the next round.

def exist(self, board, word):
    if word == "":
        return True
    m = len(board)
    n = len(board[0])
    def backtrack(x, y, word):
        if len(word) == 0:
            return True

        if x < 0 or x >= m or y < 0 or y >= n or board[x][y] != word[0]:
            return False

        t = board[x][y]
        board[x][y] = '#'
        res= backtrack(x+1, y, word[1:]) or backtrack(x, y+1, word[1:]) or backtrack(x-1, y, word[1:])
        board[x][y] = t
        return res
    for i in range(m):
        for j in range(n):
            if backtrack(i, j, word):
                return True

    return False

```

5.5.3.2 DFS+BFS

#934 Shortest Bridge, this is a very good problem because it covers both DFS and BFS, to find the shortest path of 2 island, the first step is to find one of the island and marked it as different element(2 in my case), because from the perspective of finding the same 1 in the deepest path, this is something DFS usually does, so dfs is used here to search an island thoroughly, and the same time enqueue all of the node found in dfs, this is kinda like enqueue all node in the same level(From the perspective of another island).

The second step is to use classical BFS, the node marked as 2 could be re-used as ‘visited’ set in BFS.

```

def shortestBridge(self, A):
    # dfs and mark an island to 2
    m = len(A)
    n = len(A[0])

    q = collections.deque()
    # dfs
    def backtrack(x, y):
        # end condition
        if x == m or x < 0 or y == n or y < 0 or A[x][y] != 1:
            return
        A[x][y] = 2
        q.append((x, y))
        for a, b in ((x+1, y), (x-1, y), (x, y+1), (x, y-1)):
            backtrack(a, b)

    Found = False
    for i in range(m):
        for j in range(n):
            if A[i][j] == 1:
                backtrack(i, j)
                Found = True
                break
        if Found == True:
            break

    # bfs
    dirs = [0, -1, 0, 1, 0]
    step = 0
    while q:
        qSize = len(q)
        for i in range(qSize):
            x, y = q.popleft()
            for j in range(4):
                a = x + dirs[j]
                b = y + dirs[j + 1]

                if a == m or a < 0 or b == n or b < 0 or A[a][b] == 2:
                    continue
                A[a][b] = 2
                q.append((a, b))
                step += 1

```

```

if A[a][b] == 1:
    return step

A[a][b] = 2
q.append((a, b))

step += 1

return -1

```

5.6 Binary Search

The Binary Search Algorithm is very easy, but the detail is very important, lower bound, upper bound, return mid + 1 or mid - 1. The open interval and closed interval are also a question that needs to be remember.

The template for the Binary Search is as follows:

```

# return the first value that is not less than value in [first, last)
def lower_bound(array, first, last, value):
    while first < last:
        mid = first + (last - first) // 2
        if array[mid] < value: first = mid + 1
        else: last = mid
    return first # last is also fine

```

There are 2 cases here:

First is to find lower bound, which means to find $x \geq value$ or $x > value$, the index of x is the minimum. Use $[first, last)$ interval.

Second is to find the upper bound, which is to find the $x \mid value$ or $x \leq value$, where the index of x is the largest.

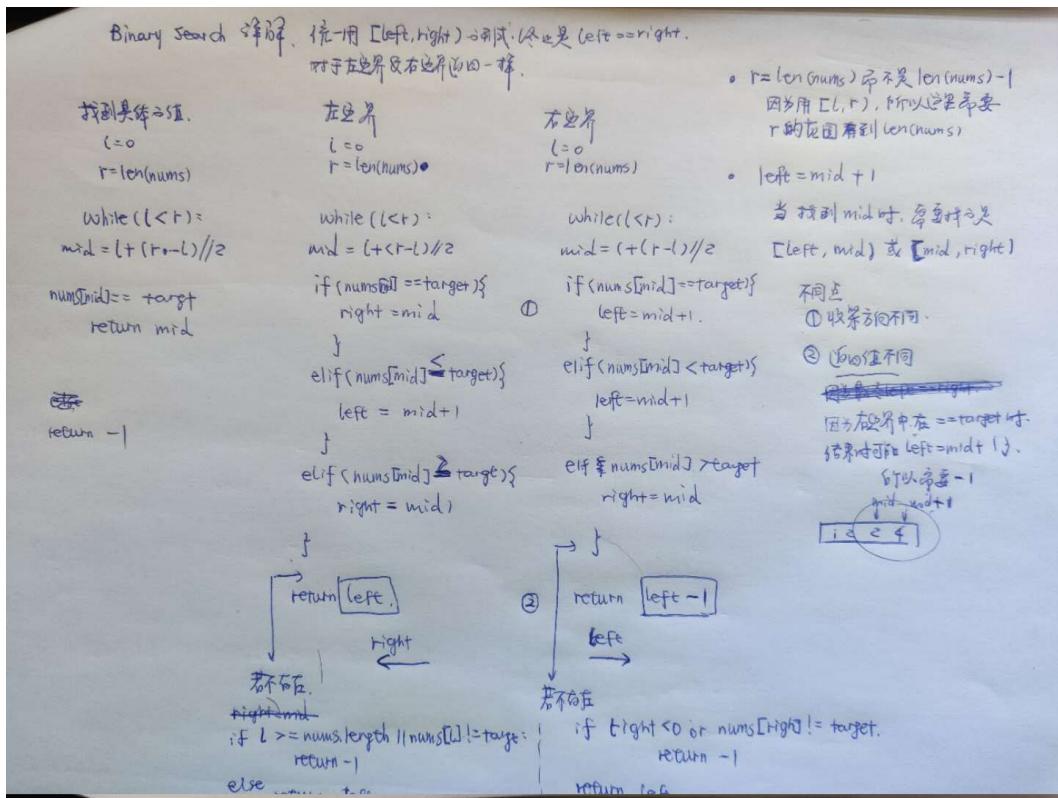


Figure 5.11: peak number

leetcode #33 want to find a value inside a rotated array, this actually could be solved by using indexof but since the problem should be solved in nlogn, then binary search should be used here. The important part is **to know which part is strictly increasing sequence**, for example, a sequence is 4,5,6,7,0,1,2. say if we want to find value 0, if mid is in 7, then we simply compare mid with r and l to see which part is sorted, obviously, 4,5,6,7 is sorted, while 7,0,1,2 is not sorted, after know which part is sorted, just simply compare it with **target to see whether target is in this fields**. if not, then shrink the mid. Remember we need to compare our mid target with the right value, so we will use closure $[l, r]$ instead of $[l, r]$ in this one.

```

def search(self, nums, target):
    """
    :type nums: List[int]
    :type target: int
    :rtype: bool
    """

    N = len(nums)
    l, r = 0, N - 1
    while l <= r:

```

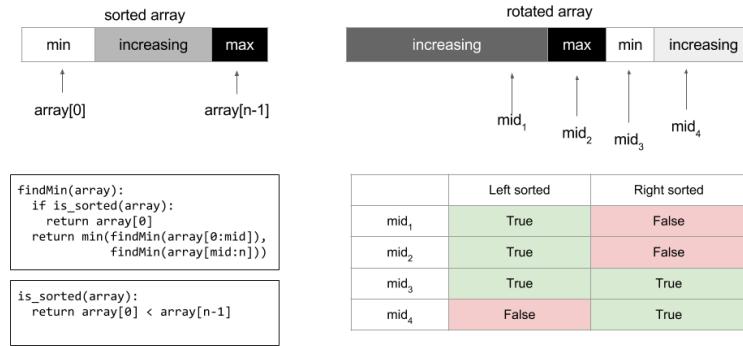
```

while l < r and nums[l] == nums[r]:
    l += 1
mid = l + (r - 1) / 2
if nums[mid] == target:
    return mid
if nums[mid] >= nums[l]:
    if nums[l] <= target < nums[mid]:
        r = mid - 1
    else:
        l = mid + 1
elif nums[mid] <= nums[r]:
    if nums[mid] < target <= nums[r]:
        l = mid + 1
    else:
        r = mid - 1
return -1

```

in leetcode #153, try to find the minimum of a rotated array, just use recursion+binary search to do that.

153. Find Minimum in Rotated Sorted Array



花花酱 LeetCode

Figure 5.12: Minimum number of Rotated array 1

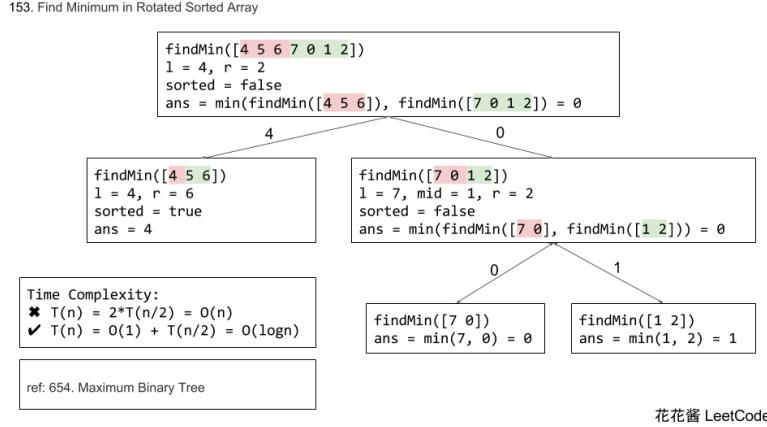


Figure 5.13: Minimum number of Rotated array 1

```

def findMin(self, nums):
    def bsMin(nums, l, r):
        # bs, end condition is l==r(only 1 element), if 2 left, l+1==r
        if l + 1 >= r:
            return min(nums[l], nums[r])

        # sorted
        if nums[l] < nums[r]:
            return nums[l]

        mid = l + (r - l) // 2
        return min(bsMin(nums, mid + 1, r), bsMin(nums, l, mid))

    l = 0
    r = len(nums) - 1
    return bsMin(nums, l, r)

```

leetcode #162 peak number, need to find a peak number in an array, use 2 pointers in this case. The process is shown below.

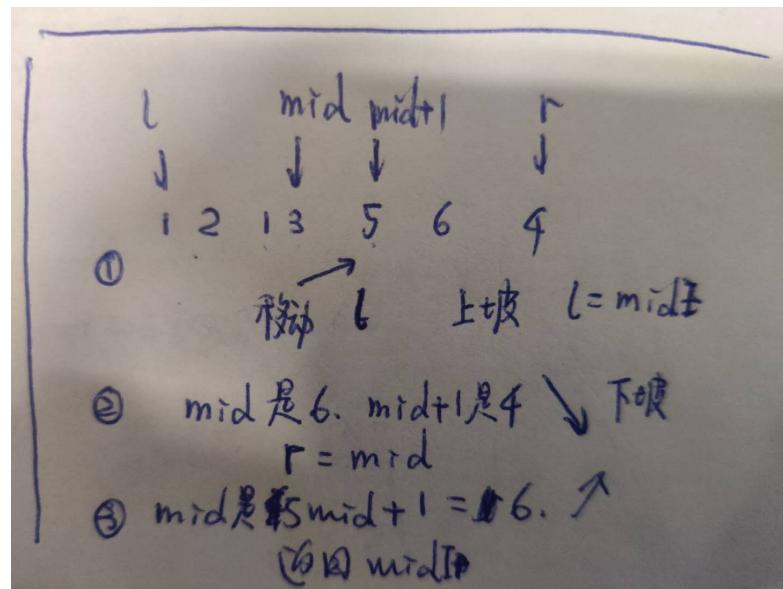


Figure 5.14: peak number

```
def findPeakElement(self, nums):
    l = 0
    r = len(nums) - 1
    while l < r:
        mid = l + (r - 1) // 2
        mid1 = mid + 1

        if nums[mid] < nums[mid1]:
            l = mid1
        else:
            r = mid

    return l
```

#69 upper bound, be careful that the l should be 1 instead of 0, the reason to find upper bound instead of lower bound is that **to find the first that is bigger than the result or say the rightest one that close to the result**

```
def mySqrt(self, x):
    l = 1
    r = x

    while l < r:
        mid = l + (r - 1) // 2
        if mid ** 2 == x:
```

```

        return mid
    elif mid ** 2 > x:
        r = mid
    else:
        l = mid + 1

    return l - 1

```

#378 Find the smallest x(lower bound Binary Search), such that there are k elements that are smaller or equal to x(upper bound).

```

def kthSmallest(self, matrix, k):
    if not matrix:
        return -1

    l = matrix[0][0]
    r = matrix[-1][-1] + 1

    while l < r:
        mid = l + (r - l) // 2

        loc = sum(bisect_right(a, mid) for a in matrix)

        if loc == k:
            r = mid
        elif loc < k:
            l = mid + 1
        else:
            r = mid

    return l

```

#668, it is a little bit different from the #378, the matrix is just the multiplication of col and row, if use the way above(bisect_right) will cause memory limit, so instead of using sum over the bisect_right, we use a function called lex, for every rows(m), compare the col with mid/i , i is every row here, we take advantage of the fact that each cell in matrix is actually just a multiplication, mid/i can calculate the boundary in this line, for example, [1,2,3],[2,4,6],[3,6,9], the first mid will be 5, then when scan the first row, n=3, $mid/1 = 5$, so in the first line there are $\min(3,5)$ 3 numbers that is smaller than the mid, for the second line, m=2, mid=5, n=3, $mid//m = 2.5$, then in this case, there are 2 numbers that is smaller than 5 (2 and 4). The following slide shows the detail reasoning.

LeetCode YouTube bilibili 花花酱 huahualeetcode 668. Kth Smallest Number in Multiplication Table

<p>Solution 1: Brute Force (MLE) Generate the table, and find k-th element Time complexity: $O(m^2n)$ Space complexity: $O(m^2n) = 0.9 \cdot 4 = 3.6$ GB</p> <p>Solution 2: Binary Search Target: the k-th element x. Find smallest x such that there are at least k elements $\leq x$ in the table</p> <pre> l = 1, r = n * n + 1 while l < r: x = l + (r - 1) // 2 if lex(x) >= k: r = x else: l = x + 1 return l </pre> <p>Time complexity: $O(m \cdot \log(m^2n))$ Space complexity: $O(1)$</p>	<p>how many #s are $\leq x$ in the table.</p> <p>Do it row by row</p> $\begin{array}{ccccccc} 1 & 2 & 3 & \dots & n^*1 \\ 2 & 4 & 6 & \dots & n^*2 \\ \dots & & & & & & \\ i & 2*i & 3*i & \dots & n*i & \leq x \\ \dots & & & & & & \\ m & 2*m & 3*m & \dots & n*m \end{array}$ <p>$1, 2, 3, \dots, n \leq x/i$ (both $/i$) $\#s \leq x = \min(x/i, n)$</p> <pre> def lex(x): # O(m) return sum(min(n, x/i)) i=1..m </pre> <p>Related Problems: k-th element LC378 <= LC668 < LC786</p>	<p>Why use $\text{lex}(x) \geq k$, not $\text{lex}(x) == k$? There are duplications! $n = 3, m = 3, k = 7$</p> $\begin{array}{ccccccc} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 3 & 6 & 9 \end{array}$ <p>Sorted: 1,2,2,3,3,4,6,6,9</p> <p>$\text{lex}'(9) = 9 \geq 7$ $\text{lex}'(6) = 8 \geq 7$ $\text{lex}'(4) = 6$</p> <p>Pruning: 1. sum is already $\geq k$, $\text{lex}(x) \geq k$ is already true</p> <p>2. x/i is already 0, i goes up x/i goes down, sum will not change after that</p>
--	--	--

<https://youtube.com/c/huahualeetcode><https://space.bilibili.com/9880352><https://zxi.mytechroad.com/blog/>

225

Figure 5.15: Kth Smallest Number in Multiplication Table (From huahua)

```

def findKthNumber(self, m, n, k):
    l = 1
    r = m*n+1

    def lex(mid, m, n):
        count = 0
        for i in range(1, m+1):
            count += min(n, mid // i)
        return count

    while l < r:
        mid = l + (r - 1) // 2
        loc = lex(mid, m, n)

        if loc >= k:
            r = mid
        else:
            l = mid + 1

    return l

```

#786, a hard question, should take advantage of the property of the matrix, for each row, found the first col that is $A[i]/A[j] \leq m$, in other words, $A[i] < A[j] * m$, so the rest will all smaller. The detail is shown below.

LeetCode YouTube 花花酱 huahualeetcode

786. K-th Smallest Prime Fraction

			smallest
1/2	1/3	1/5	
-	2/3	2/5	A = [1, 2, 3, 5], K = 3
-	-	3/5	def: M[i][j] = A[i] / A[j] 0 <= i <= n - 1, i < j < n - 1 M[i][j] < M[i][j - 1] M[i][j] > M[i][j - 1]

Solution 0: Brute Force / Sorting TLE

Time complexity: $O(n^2 \log n^2)$
Space complexity: $O(n^2)$

Solution 1: Binary Search Accepted

Related Problems: LC 378/719

Search m, $0 < m < 1$
Such that there are exact K pairs of (i,j) that $A[i] / A[j] \leq m$

Create a virtual matrix to speed up the search process

Time complexity: $O(\log(\max(n)^2))$ TLE
 $O(\log(\max(n)))$
Space complexity: $O(1)$

if $M[i][j] > m \Rightarrow M[i + 1][j] \geq m$

I	r	m	$A[i] / A[j] \leq m$	total
0	1	0.5	1/2, 1/3, 1/5, 2/5	4 > K
0	0.5	0.25	1/5	1 < K
0.25	0.5	0.375	1/3, 1/5	2 < K
0.375	0.5	0.4375	1/3, 1/5, 2/5	3 = K

<http://zxi.mytechroad.com/blog/>

Figure 5.16: K-th Smallest Prime Fraction

```
def kthSmallestPrimeFraction(self, A, K):
    l = 0.0
    r = 1.0
    n = len(A)

    while l < r:
        mid = (l + r) / 2
        max_f = 0.0
        total = 0
        j = 1
        for i in range(n - 1):
            while j < n and A[i] > A[j] * mid:
                j += 1
            if j == n:
                break
            total += n - j
            f = A[i] / A[j]
            if f > max_f:
                p, q, max_f = i, j, f

        if total == K:
            return [A[p], A[q]]
        elif total > K:
            r = mid
        else:
            l = mid
```

```

else:
    l = mid

return []

```

5.7 Dynamic Programming

Normally the problem of DP is the problem of **finding maximum or minimum solutions**, that is to say, to find optimal solution in operation research, for example, longest increasing sub-sequence, minimum edit distance. To find the extremum, the core is **exhaustive method**, a little different from pure exhaustive method, DP related questions should be optimized using DP table, after the dp table, the dynamic transition function and optimal sub structure is the key to the answer. So the most difficult part is to write down the **dynamic transition function**. The process to write down DTF is as follows:

- base case
- know the state space
- choices
- defined the dp matrix

```

1 # 初始化 base case
2 dp[0][0][...] = base
3 # 进行状态转移
4 for 状态1 in 状态1的所有取值：
5     for 状态2 in 状态2的所有取值：
6         for ...
7             dp[状态1][状态2][...] = 求最值(选择1, 选择2...)

```

Figure 5.17: DP vague template

5.7.1 Simple Example: Fibonacci

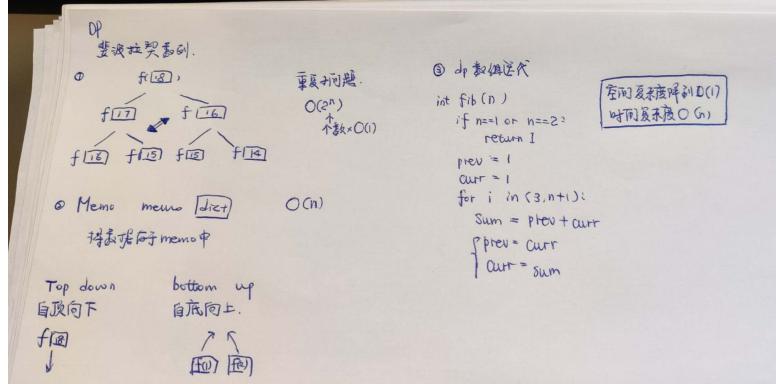


Figure 5.18: DP Example

5.7.2 Coin Change

#322 You are given coins of different denominations and a total amount of money amount. Write a function to compute the fewest number of coins that you need to make up that amount.

The base case here is when amount is 0 then we only need 0 coin. State is the target amount, and choice is the coins you have, the dp array is defined as: input n aim amount, the dp array will return the fewest amount of coins

LeetCode
YouTube
花花酱
huahualeetcode
322. Coin Change

Solution1: DP

Def: $dp[i][j]$: min coins to make up j amount using first i types of coins.

Init: $dp[-1][0] = 0$, $dp[-1][j] = \infty$

Transition: $dp[i][j] = \min(dp[i][j], dp[i - 1][j - k * coin_i] + k)$
 $= \min(dp[i][j], dp[i][j - coin_i] + 1)$

Ans: $dp[n-1][amount]$

Time complexity: $O(n * amount^2) \rightarrow O(n * amount)$

Space complexity: $O(n * amount) \rightarrow O(amount)$

num	dp[0]	dp[1]	dp[2]	dp[3]	dp[4]	dp[5]	dp[6]	dp[7]	dp[8]	dp[9]	dp[10]	dp[11]
-	0	∞										
1	0	1	2	3	4	5	6	7	8	9	10	11
2	0	1	1	2	2	3	3	4	4	5	5	6
5	0	1	1	2	2	1	2	2	3	3	2	3

<http://zxi.mytechroad.com/blog/>

Figure 5.19: Coin Change

It is noticeable that defined the space of dp as amount + 1 instead of amount, because 0 to amount has the number of amount + 1.

```
def coinChange(self, coins, amount):
```

```

dp = [float('inf')] * (amount + 1)

dp[0] = 0
for i in range(amount+1):
    for coin in coins:
        if coin > i:
            continue
        dp[i] = min(dp[i], dp[i - coin] + 1)

return -1 if dp[-1] == float('inf') else dp[-1]

```

For #377, it is the same the only difference is that the base case $dp[0]$ is 1, which means there is 1 way to sum up to target 0, and since it is not the case of getting the min, then sum up all of the possibilities.(The dp defines as given the target index, return the number of possibilities that could sum up to target index)

```

class Solution(object):
    def combinationSum4(self, nums, target):
        """
        :type nums: List[int]
        :type target: int
        :rtype: int
        """
        dp = [0] * (target+1)
        dp[0] = 1
        for i in range(target+1):
            for num in nums:
                if i - num >= 0:
                    dp[i] += dp[i - num]
        return dp[-1]

```

5.7.3 Longest Increasing Subsequence (LIS) $O(n^2)$

Problem statement: given a non sorted array, find the longest increasing subsequence. be careful about the concept of subsequence and substring. substring should be continuous while subsequence does not have to be continuous.

define dp array, the $dp[i]$ array means the longest subsequence of a certain number $nums[i]$, the dp array should be initialized as 1 since itself count as 1. The process is as follows:

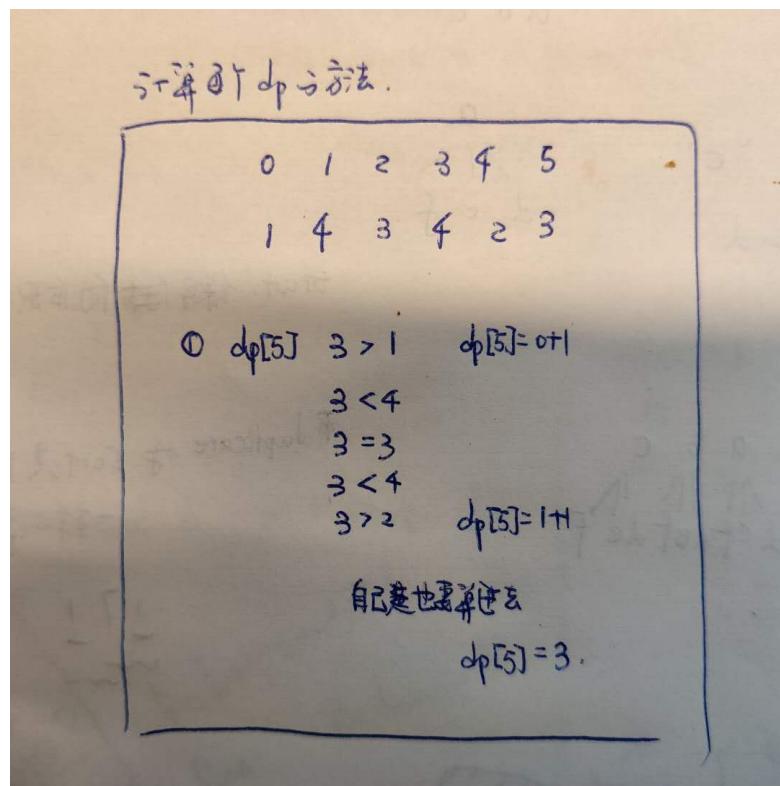


Figure 5.20: LIS

Remember that in the process shown above, $dp[5]$ should be $1 + 1$ or $dp[5] = 2 + 1$, the 1 and 2 actually is the result from $dp[0]$ and $dp[4]$. That explains the part of $dp[i] = \max(dp[i], dp[j] + 1)$. #310

```
def lengthOfLIS(self, nums):
    """
    :type nums: List[int]
    :rtype: int
    """
    if not nums:
        return 0

    l = len(nums)
    dp = [1] * l
    for i in range(1, l):
        for j in range(i):
            if nums[j] < nums[i]:
                dp[i] = max(dp[i], dp[j] + 1)

    return max(dp)
```

#673 is to know the number of LIS, when found the longest one, update LISN[i], if found the same, then sum them up. for example (1,3,4,7) and (1,3,5,7)

```
def findNumberOfLIS(self, nums):
    """
    :type nums: List[int]
    :rtype: int
    """

    if not nums:
        return 0
    l = len(nums)
    LIS = [1] * l
    LISN = [1] * l
    for i in range(1, l):
        for j in range(i):
            if nums[i] > nums[j]:
                if LIS[j] + 1 > LIS[i]:
                    LIS[i] = LIS[j] + 1
                    LISN[i] = LISN[j]
                elif LIS[j] + 1 == LIS[i]:
                    LISN[i] += LISN[j]
    total = 0
    a = max(LIS)
    for i in range(l):
        if LIS[i] == a:
            total += LISN[i]

    return total
```

Follow up, for #673, to find the Number of Longest Increasing Subsequence, then only compare to the one before the current element i.

```
def findLengthOfLCIS(self, nums):
    if not nums:
        return 0
    dp = [1] * len(nums)
    for i in range(1, len(nums)):
        if nums[i] > nums[i-1]:
            dp[i] = max(dp[i], dp[i-1]+1)
    # print(dp)
    return max(dp)
```

5.7.4 Maximum Subarray

Problem: given an array and return the maximum subarray (sum up), this sumarray should be continuous. The dp array should be defined as dp[i] stands for the maximum

subarray. The transition function is reasoning like this: if we know the $dp[i-1]$, then $dp[i]$ will be known by taking 2 choices, take the next element or not.

Then the transition is:

The algorithm could be written as

```
def maxSubArray(self, nums):
    if not nums:
        return 0

    dp = [0] * len(nums)
    dp[0] = nums[0]
    for i in range(1, len(nums)):
        dp[i] = max(dp[i], dp[i-1] + nums[i])

    return max(dp)
```

Notice that $dp[i]$ only relies on $dp[i-1]$, then we can shrink the space to $O(1)$

```
def maxSubArray(self, nums):
    if not nums:
        return 0

    dp0 = nums[0]
    dp1 = 0
    max1 = dp0
    for i in range(len(nums)):
        dp1 = max(nums[i], dp0 + nums[i])
        dp0 = dp1
        max1 = max(max1, dp0)
    return max1
```

Notice that

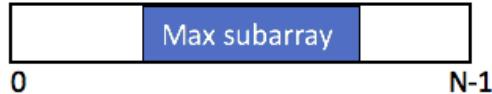
```
def maxSubarraySumCircular(self, A):
    """
    :type A: List[int]
    :rtype: int
    """

    maxSum, curMax= A[0], 0
    for a in A:
        curMax = max(curMax + a, a)
        maxSum = max(maxSum, curMax)

    return maxSum
```

Now change this problem to a circular list, then there are 2 cases that should be taken into account. 893 Maximum Sum Circular Subarray

Case 1: max subarray is not circular.



Case 2: max subarray is circular.

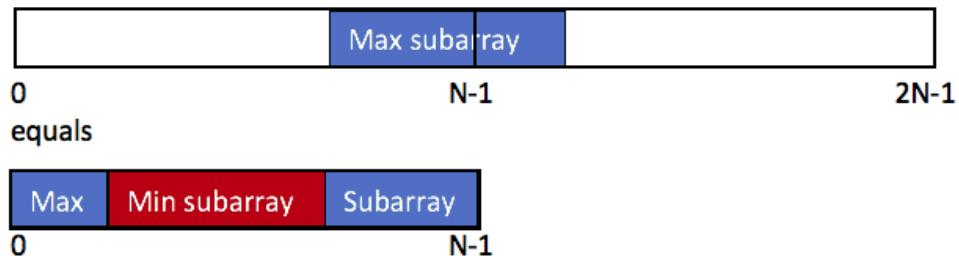


Figure 5.21: Maximum Sum Circular Subarray

```
def maxSubarraySumCircular(self, A):
    """
    :type A: List[int]
    :rtype: int
    """
    if not A:
        return 0

    total, maxSum, curMax, minSum, curMin = 0, A[0], 0, A[0], 0
    for a in A:
        curMax = max(curMax + a, a)
        maxSum = max(maxSum, curMax)
        curMin = min(curMin + a, a)
        minSum = min(minSum, curMin)
        total += a
    return max(total - minSum, maxSum) if maxSum > 0 else maxSum
```

If all numbers are negative, $\text{maxSum} = \max(A)$ and $\text{minSum} = \text{sum}(A)$. In this case, $\max(\text{maxSum}, \text{total} - \text{minSum}) = 0$, which means the sum of an empty subarray. According to the decription, We need to return the $\max(A)$, instead of sum of am empty subarray. So we return the maxSum to handle this corner case.

5.7.5 01 Knapsack problem

Problem description: given a knapsack that can load w weight and N things, every object has two properties, object i is $wt[i]$ and the value is $val[i]$, what's the most valuable things combination you could have given this knapsack?

First of all, two points must be clarified, **status** and **choice**. As long as the capacity of a few things and a backpack is limited, a backpack problem is formed. When it comes to status, it's easy to think about, for each item, what can you choose? The choice is to "pack in a backpack" or "do not pack in a backpack".

#474. Ones and Zeroes. 2 status inside: number of m and number of n. And the choices is form this word or not. The dp array here will be defined as given m zeros and n ones, what is the maximum number of strings that can be formed.

```
def findMaxForm(self, strs, m, n):
    dp = [[0 for _ in range(n+1)] for _ in range(m+1)]

    for s in strs:
        # count zeros and ones
        zeros, ones = 0, 0
        for c in s:
            if c == '0':
                zeros += 1
            if c == '1':
                ones += 1

        # status and select
        for i in range(m, zeros-1, -1):
            for j in range(n, ones-1, -1):
                # choice: form this word or not
                # if form this word, the number of zeros in this word will be deducted
                dp[i][j] = max(dp[i - zeros][j - ones] + 1, dp[i][j])

    return dp[m][n]
```

5.7.6 Subset Knapsack problem

#416, given an non empty array with only positive numbers, find if the array can be partitioned into two subsets such that the sum of elements in both subsets is equal.

This problem is also a Knapsack problem, given a sum/2 weight knapsack with n number, is there a way to fulfill this knapsack. The dp array could be defined as $dp[i][j]$, where i denotes the first i numbers, and j denotes the volume of the knapsack. If $dp[i][j]$ is true then it can be fulfilled with the first i numbers. The states are those number and choices are choose the i number or not.

transition function then becomes:

- if $\text{num}[i]$ is not included, then $\text{dp}[i][j] = \text{dp}[i-1][j]$, same as the last one
- if $\text{num}[i]$ is included, then $\text{dp}[i][j] = \text{dp}[i-1][j-\text{nums}[i]]$

如果不把 $\text{nums}[i]$ 算入子集, 或者说你不把这第 i 个物品装入背包, 那么是否能够恰好装满背包, 取决于上一个状态 $\text{dp}[i-1][j]$, 继承之前的结果。

如果把 $\text{nums}[i]$ 算入子集, 或者说你把这第 i 个物品装入了背包, 那么是否能够恰好装满背包, 取决于状态 $\text{dp}[i-1][j-\text{nums}[i-1]]$ 。

Figure 5.22: Knapsack

```
def canPartition(self, nums):
    nums.sort()
    sum1 = sum(nums)
    aim = sum1/2
    if sum1%2 != 0:
        return False

    dp = [[0 for _ in range(aim+1)] for _ in range(len(nums) + 1)]
    dp[0][0] = True

    for i in range(1, len(nums) + 1):
        for j in range(1, aim+1):
            if nums[i - 1] > j:
                # not enough j, can't put the i number
                dp[i][j] = dp[i-1][j]
            else:
                # put in or not put in
                dp[i][j] = dp[i-1][j] or dp[i-1][j-nums[i-1]]

    return dp[-1][-1]
```

Notice that the $\text{dp}[i][j]$ only depends on the last state $\text{dp}[i-1][...]$, then we could do a space shrink, and j should be traversed backwards. The current position $\text{dp}[i][j]$ before actually is replied on the last state $+ \text{nums}[i]$, if we do it forward, then $\text{dp}[i-2]$ will influence $\text{dp}[i-1]$, and then $\text{dp}[i-1]$ will also be influenced, which means a $\text{nums}[i]$ is used several times.

```
def canPartition(self, nums):
    nums.sort()
    sum1 = sum(nums)
```

```

aim = sum1/2
if sum1%2 != 0:
    return False

dp = [0 for _ in range(aim+1)]
dp[0] = True

for i in range(1, len(nums) + 1):
    for j in range(aim, -1, -1):
        if nums[i - 1] <= j:
            # put in or not put in
            dp[j] = dp[j] or dp[j-nums[i-1]]

return dp[-1]

```

5.7.7 Complete Knapsack Problem

The complete Knapsack problem is similar to the problems mentioned before, the only difference is that we could used the things or numbers infinite times.

#518, CoinChange 2, amount 5, coins=[1,2,5]. count how many times it has to sum up to 5 using the coins.

- states: amount and coins
- dp defined as dp[num][amount], using only the first num of coin, the all possibilities.
- base cases: dp[0][...] =0, dp[...][0] = 1

```

def change(self, amount, coins):
    if not coins and amount == 0:
        return 1

    dp = [[0 for i in range(amount+1)] for j in range(len(coins) + 1)]
    for i in range(len(coins)+1):
        dp[i][0] = 1

    for i in range(1, len(coins) + 1):
        for j in range(1, amount+1):
            # put it or not
            if j >= coins[i-1]:
                dp[i][j] = dp[i][j-coins[i-1]] + dp[i-1][j]
            else:
                dp[i][j] = dp[i-1][j]

    return dp[-1][-1]

```

```

def change(self, amount, coins):
    if not coins and amount == 0:
        return 1

    dp = [0 for i in range(amount+1)]
    dp[0] = 1

    for i in range(1, len(coins) + 1):
        for j in range(1, amount+1):
            # put it or not
            if j >= coins[i-1]:
                dp[j] = dp[j-coins[i-1]] + dp[j]

    return dp[-1]

```

5.7.8 Edit Distance

Given 2 string s1 and s2, calculate the fewest step that transform s1 to s2. The operations are insert, delete and replace. (Actually there is a hidden operation, that is skip)

To solve 2 strings problem, especially in DP, normally we will use 2 pointers. The pseudocode will be like

```

if s1[i] == s2[j]:
    skip
    i, j move forward
else:
    3 operations:
        insert: dp(i,j-1): s1 insert, then j-1 continue compare
        delete: dp(i-1,j): delete i and continue compare
        replace: dp(i-1,j-1): replace, but this position still here, move i and j forward

```

Initialization: if one string is empty and the other is not, then the min edit distance is the length of the non empty string.

```

def minDistance(self, s1, s2):
    n1 = len(s1)
    n2 = len(s2)
    dp = [[0 for i in range(n2 + 1)] for j in range(n1 + 1)]

    # base case
    for i in range(n1+1):
        dp[i][0] = i
    for j in range(n2+1):
        dp[0][j] = j

```

```

for i in range(1, n1+1):
    for j in range(1, n2+1):
        if s1[i-1] == s2[j-1]:
            dp[i][j] = dp[i-1][j-1]
        else:
            dp[i][j] = min(dp[i-1][j], dp[i][j-1], dp[i-1][j-1]) + 1

return dp[-1][-1]

```

5.7.9 Longest Common Subsequence(LCS)

#1143 Given two strings text1 and text2, return the length of their longest common subsequence. Using 2 pointers, if $s1[i] == s2[j]$, then this letter is definitely in LCS, otherwise, at least one of the letter inside s1 or s2 should be not in lcs. take the max one.

```

def longestCommonSubsequence(self, s1, s2):
    n1 = len(s1)
    n2 = len(s2)

    dp = [[0 for i in range(n2+1)] for j in range(n1+1)]

    # initialization one is empty then LCS is 0
    # dp[i][0] = 0   dp[0][j] = 0
    for i in range(1, n1+1):
        for j in range(1, n2+1):
            # found common
            if s1[i-1] == s2[j-1]:
                dp[i][j] = dp[i-1][j-1] + 1
            else:
                dp[i][j] = max(dp[i-1][j], dp[i][j-1])

    return dp[-1][-1]

```

5.7.10 Longest Palindromic Subsequence

#516 Given a string s, return the Longest Palindromic Subsequence. The template is similar to LCS, dp array defined as $dp[i][j]$: the LPS($nums[i]$ $nums[j]$). The reasoning of the transition function is illustrated below.

The 2 pointers i and j will go from 2 directions. i will go forward and j will go backward.

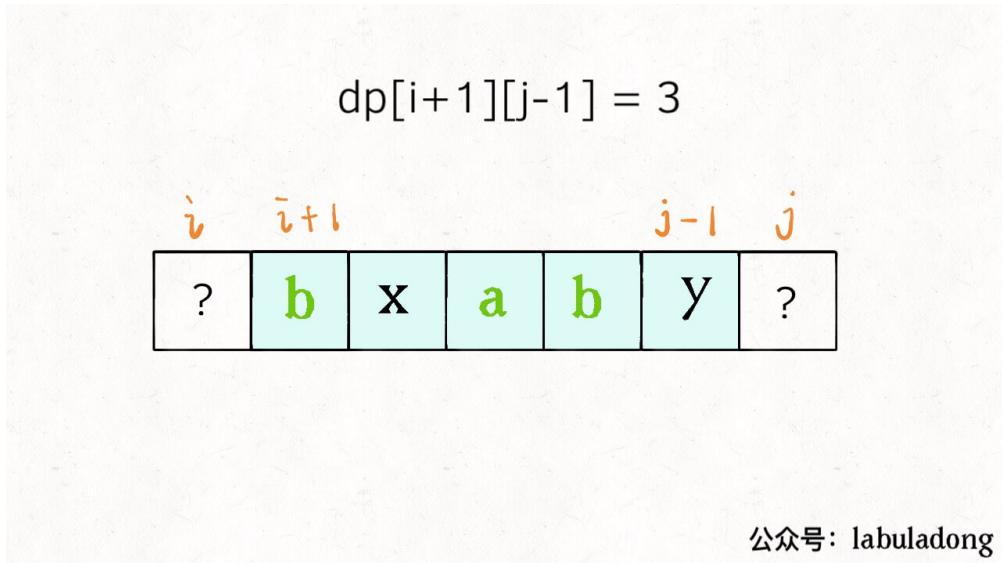


Figure 5.23: Ex1

If the two nums are the same, simply plus 2.

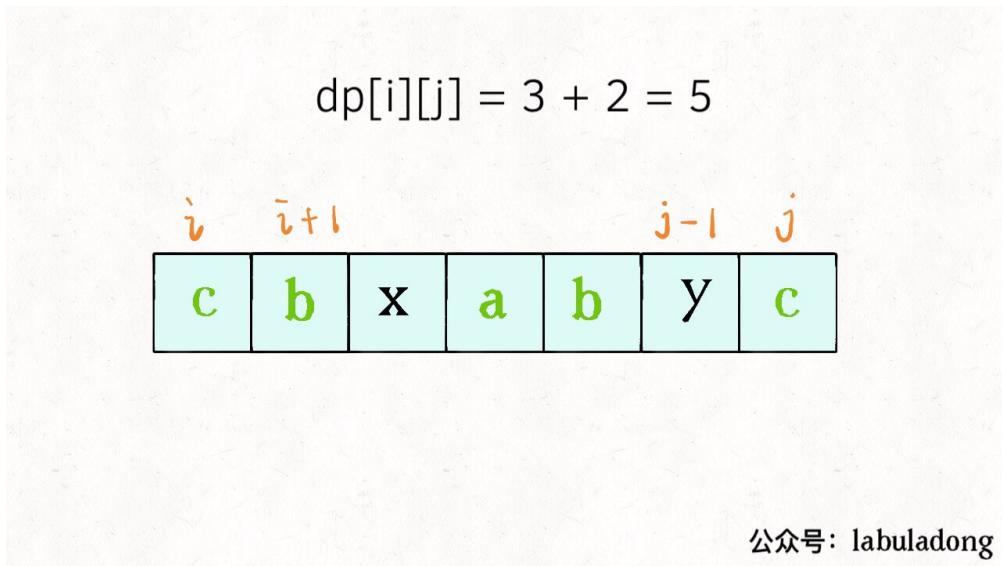


Figure 5.24: Ex2: when 2 nums equal

If not the same, take the maximum of one of them.

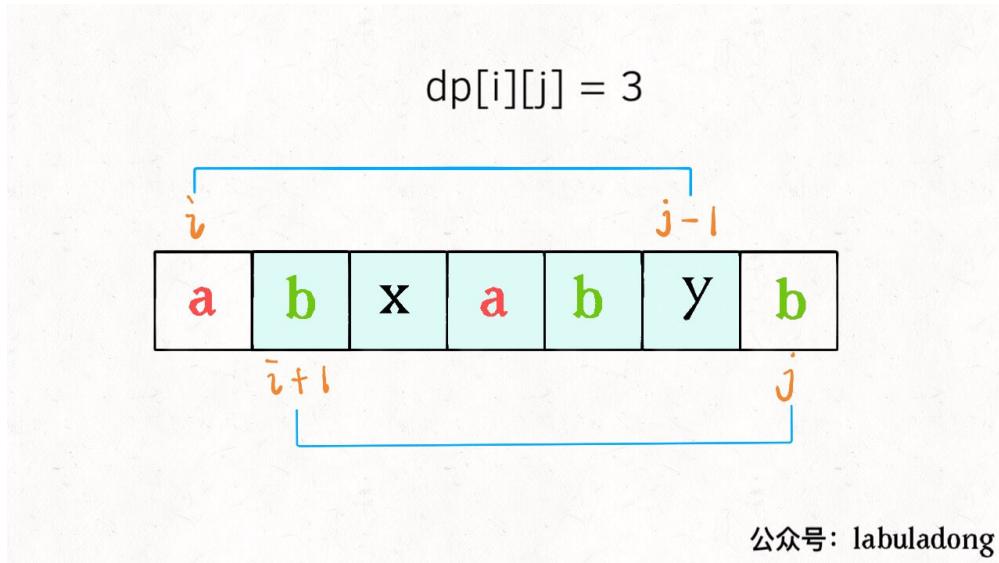


Figure 5.25: Ex3: When 2 nums not the same

Base Cases: If $i == j$, then only 1 elements. when $i \neq j$, there is no elements.

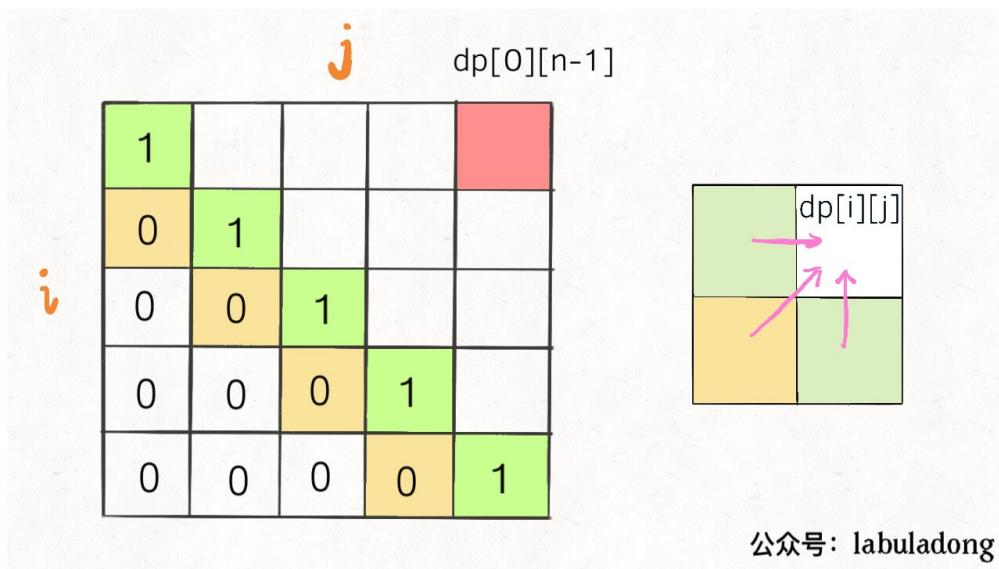


Figure 5.26: Base Cases

```
def longestPalindromeSubseq(self, s):
    n = len(s)

    dp = [[0 for _ in range(n)] for _ in range(n)]
    # base cases
```

```

for i in range(n):
    for j in range(n):
        if i == j:
            dp[i][j] = 1
        # to make sure dp[i][j] can be updated
        # i should be start from n-1
    for i in range(n-1,-1,-1):
        for j in range(i+1, n):
            if s[i] == s[j]:
                dp[i][j] = dp[i+1][j-1] + 2
            else:
                dp[i][j] = max(dp[i+1][j], dp[i][j-1])

return dp[0][n-1]

```

5.7.11 Super Egg Drop

#887, You are given K eggs, and you have access to a building with N floors from 1 to N. You know that there exists a floor F with $0 \leq F \leq N$ such that any egg dropped at a floor higher than F will break, and any egg dropped at or below floor F will not break. Your goal is to know with certainty what the value of F is. What is the **minimum** number of moves that you need to know with certainty what F is

Actually this problem could be solved using Binary Search, but the problem is that the number of eggs is given, which means you can't do experiments whatever times you want.

- states: Number of eggs K and Number of floor N.
- Choices: which floor to drop the egg. The result after choosing this floor i also has 2 situations, the egg breaks or not. If the egg breaks, K-1, And the search space should be transformed from [1..N] to [1..i-1], if not, K. the space is then [i+1..N]

$$dp(K, N) = \min_{0 <= i <= N} \{ \max\{dp(K - 1, i - 1), dp(K, N - i)\} + 1 \}$$

Figure 5.27: Drop Egg Transition function

```

def superEggDrop(self, K, N):

    memo = collections.defaultdict()
    def dp(K, N):
        # base case

```

```

if K == 1: return N
if N == 0: return 0
#
if (K, N) in memo:
    return memo[(K, N)]

res = float('INF')
#
for i in range(1, N + 1):
    res = min(res,
               max(
                   dp(K, N - i),
                   dp(K - 1, i - 1)
               ) + 1
    )
#
memo[(K, N)] = res
return res

return dp(K, N)

```

This could be optimized with binary search:

```

memo = dict()
def dp(K, N):
    if K == 1: return N
    if N == 0: return 0
    if (K, N) in memo:
        return memo[(K, N)]

    # for 1 <= i <= N:
    #     res = min(res,
    #                max(
    #                    dp(K - 1, i - 1),
    #                    dp(K, N - i)
    #                ) + 1
    #     )

    res = float('INF')
    #
    lo, hi = 1, N
    while lo <= hi:
        mid = (lo + hi) // 2

```

```

broken = dp(K - 1, mid - 1) #
not_broken = dp(K, N - mid) #
# res = min(max() + 1)
if broken > not_broken:
    hi = mid - 1
    res = min(res, broken + 1)
else:
    lo = mid + 1
    res = min(res, not_broken + 1)

memo[(K, N)] = res
return res

return dp(K, N)

```

5.7.12 Burst Balloons

#312 Given n balloons, indexed from 0 to n-1. Each balloon is painted with a number on it represented by array nums. You are asked to burst all the balloons. If you burst balloon i you will get $\text{nums}[\text{left}] * \text{nums}[i] * \text{nums}[\text{right}]$ coins. Here left and right are adjacent indices of i. After the burst, the left and right then becomes adjacent. Find the maximum coins you can collect by bursting the balloons wisely.

312. 戳气球

难度 困难 271 收藏 分享 切换为英文 关注 反馈

有 n 个气球，编号为 0 到 $n-1$ ，每个气球上都标有一个数字，这些数字存在数组 `nums` 中。

现在要求你戳破所有的气球。每当你戳破一个气球 i 时，你可以获得 $nums[left] * nums[i] * nums[right]$ 个硬币。这里的 `left` 和 `right` 代表和 i 相邻的两个气球的序号。注意当你戳破了气球 i 后，气球 `left` 和气球 `right` 就变成了相邻的气球。

求所能获得硬币的最大数量。

说明:

- 你可以假设 $nums[-1] = nums[n] = 1$ ，但注意它们不是真实存在的所以并不能被戳破。
- $0 \leq n \leq 500, 0 \leq nums[i] \leq 100$

示例:

```
输入: [3,1,5,8]
输出: 167
解释: nums = [3,1,5,8] --> [3,5,8] --> [3,8] --> [8] --> []
      coins = 3*1*5 + 3*5*8 + 1*3*8 + 1*8*1 = 167
```

Figure 5.28: Burst Balloons Example

This is a hard problem simply because the sub problems are not independent. the score of bursting $nums[i]$ is related to $nums[i-1]$ and $nums[i+1]$. The first is to define dp array. To simplify this problem, add 2 element $nums[-1] = 1$ and $nums[n] = 1$.

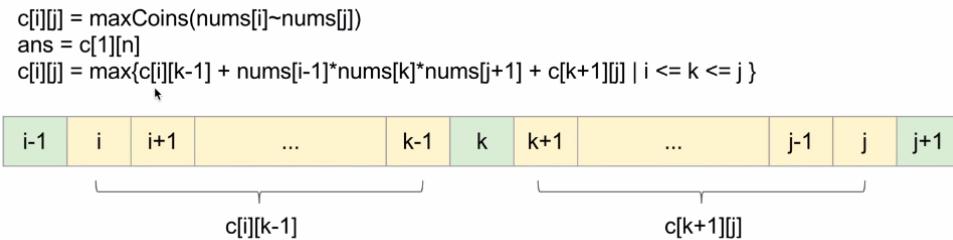


Figure 5.29: Burst Balloons TTransition Function

```
def maxCoins(self, nums):

    n = len(nums)
    nums.insert(0, 1)
    nums.append(1)
```

```

# dp[i][j] means maxCoin(nums[i]^nums[j])
c = [[0 for _ in range(n+2)] for _ in range(n+2)]

for l in range(1, n+1):
    # i could be in 1^n
    for i in range(1, n-l+2):
        j = i+l-1
        # k in between
        for k in range(i, j+1):
            c[i][j] = max(c[i][j], c[i][k-1] + nums[i-1]*nums[k]*nums[j+1] + c[i][k])

return c[1][n]

```

5.7.13 Stock Buying Selling Problem

Say you have an array for which the i th element is the price of a given stock on day i . If you were only permitted to complete at most one transaction (i.e., buy one and sell one share of the stock), design an algorithm to find the maximum profit. A more general problem is given a list of prices, but the transition k could be more than 1, but in this case, only 1 transaction is allowed.

- States: The first is days, the second is the maximum transaction time k (in this case is 1). The third state is whether you have stock in your hand.
- Choices: Buy, Sell and hold(do nothing)

Base Case: Since we want to make this kind of problems more general to be solved, in the base case, therefore the detail will be elaborated. 1 in the 3 dimension means have stock, 0 stands for no stock

- $dp[-1][k][0] = 0$: i starts from 0. -1 means haven't started, initialize the profit to 0.
- $dp[-1][k][1] = \text{float}('inf')$: this situation is not possible, set it to inf because it is impossible to hold stock before the first day
- $dp[i][0][0] = 0$. Not possible to buy or sell.
- $dp[i][0][1] = \text{float}('inf')$: not possible

So the dp array could be defined as $dp[i][k][0 \text{ or } 1]$, i is the day, k is the transaction time. Then it is just to compare 3 different choices with max. For example, $dp[3][2][1]$ means on the third day, I have stock, and I could only transact 2 times at most.

The result we want to know is $dp[n-1][K][0]$.

```

1 dp[i][k][0] = max(dp[i-1][k][0], dp[i-1][k][1] + prices[i])
2             max( 选择 rest , 选择 sell )
3
4 解释：今天我没有持有股票，有两种可能：
5 要么是我昨天就没有持有，然后今天选择 rest，所以我今天还是没有持有；
6 要么是我昨天持有股票，但是今天我 sell 了，所以我今天没有持有股票了。
7
8 dp[i][k][1] = max(dp[i-1][k][1], dp[i-1][k-1][0] - prices[i])
9             max( 选择 rest , 选择 buy )
10
11 解释：今天我持有股票，有两种可能：
12 要么我昨天就持有股票，然后今天选择 rest，所以我今天还持有股票；
13 要么我昨天本没有持有，但今天我选择 buy，所以今天我就持有股票了。

```

Figure 5.30: Stock buy sell psudocode

```

# 121, k = 1:

def maxProfit(self, prices):
    """
    :type prices: List[int]
    :rtype: int
    """

    if not prices:
        return 0
    n = len(prices)
    dp = [[0 for _ in range(2)] for _ in range(n)]

    for i in range(n):
        if i-1 == -1:
            dp[i][0] = 0
            dp[i][1] = -prices[0]
            continue
        #dp[i][1][0] = max(dp[i-1][1][0], dp[i-1][1][1] + prices[i])
        #dp[i][1][1] = max(dp[i-1][1][1], dp[i-1][0][0] - prices[i])
        #           = max(dp[i-1][1][1], -prices[i])
        dp[i][0] = max(dp[i-1][0], dp[i-1][1] + prices[i])
        dp[i][1] = max(dp[i-1][1], -prices[i])

    return dp[n-1][0]

```

By shrinking the space we can get:

```

def maxProfit(self, prices):

    if not prices:

```

```

        return 0
n = len(prices)
dp_0 = 0
dp_1 = float('-inf')
for i in range(n):
    dp_0 = max(dp_0, dp_1 + prices[i])
    dp_1 = max(dp_1, -prices[i])

return dp_0

#122

def maxProfit(self, prices):
    """
    :type prices: List[int]
    :rtype: int
    """
    n = len(prices)
    if not prices:
        return 0

    #dp=[[0 for i in range(2)] for j in range(n)]

    dp_0 = 0
    dp_1 = -prices[0]
    for i in range(n):
        temp = dp_0
        dp_0 = max(dp_0, dp_1 + prices[i])
        dp_1 = max(dp_1, temp - prices[i])

    return dp_0

#309

def maxProfit(self, prices):
    if not prices:
        return 0

    n = len(prices)
    dp = [[0 for _ in range(2)] for _ in range(n)]

    for i in range(n):
        if i - 1 == -1:

```

```

dp[i][0] = 0
dp[i][1] = -prices[0]
continue
if i - 2 == -2:
    dp[i][0] = 0
    dp[i][1] = -prices[0]
    continue
dp[i][0] = max(dp[i-1][0], dp[i-1][1] + prices[i])
dp[i][1] = max(dp[i-1][1], dp[i-2][0] - prices[i])
return dp[n-1][0]

def maxProfit(self, prices):
    if not prices:
        return 0

    n = len(prices)

    dp_0 = 0
    dp_1 = -prices[0]
    dp_pre = 0
    for i in range(n):
        temp = dp_0
        dp_0 = max(dp_0, dp_1 + prices[i])
        dp_1 = max(dp_1, dp_pre - prices[i])
        dp_pre = temp
    return dp_0

#714

def maxProfit(self, prices, fee):
    n = len(prices)
    if not prices:
        return 0

    dp_0, dp_1 = 0, float('-inf')
    for i in range(n):
        temp = dp_0
        dp_0 = max(dp_0, dp_1 + prices[i] - fee)
        dp_1 = max(dp_1, temp - prices[i])
    return dp_0

#123

def maxProfit(self, prices):

```

```

if not prices:
    return 0
n = len(prices)
dp = [[[0 for i in range(2)] for j in range(2+1)] for k in range(n)]

for i in range(n):
    for k in range(1, 3):
        if i == 0:
            dp[i][k][0] = 0
            dp[i][k][1] = -prices[0]
        continue

        dp[i][k][0] = max(dp[i-1][k][0], dp[i-1][k][1] + prices[i]);
        dp[i][k][1] = max(dp[i-1][k][1], dp[i-1][k-1][0] - prices[i]);
return dp[n-1][2][0]

#188

def maxProfit(self, k, prices):
    if not prices:
        return 0
    n = len(prices)

    max_k = k
    if max_k > n / 2:
        return self.maxProfit_inf(prices)
    dp = [[[0 for _ in range(2)] for _ in range(k+1)] for _ in range(n)]
    for i in range(n):
        for k in range(1, max_k+1):
            if i == 0:
                dp[i][k][0] = 0
                dp[i][k][1] = -prices[0]
            continue
            dp[i][k][0] = max(dp[i-1][k][0], dp[i-1][k][1] + prices[i])
            dp[i][k][1] = max(dp[i-1][k][1], dp[i-1][k-1][0] - prices[i])
    return dp[n-1][k][0]

def maxProfit_inf(self, prices):
    n = len(prices)
    dp_0=0
    dp_1=-prices[0]
    for i in range(n):
        temp = dp_0
        dp_0 = max(dp_0, dp_1 + prices[i])

```

```

        dp_1 = max(dp_1, temp - prices[i])
    return dp_0

```

5.7.14 House Robber

You are a professional robber planning to rob houses along a street. Each house has a certain amount of money stashed, the only constraint stopping you from robbing each of them is that adjacent houses have security system connected and it will automatically contact the police if two adjacent houses were broken into on the same night.

- states: the index of the current house(element)
- choices: rob or not

dp array is defined as: $dp[i]=x$, with only 0-i element, what is the maximum profit.

#198

```

def rob(self, nums):
    if not nums:
        return 0
    n = len(nums)
    dp = [0 for i in range(n)]
    if n == 1:
        return nums[0]
    if n == 2:
        return max(nums[0], nums[1])
    # basecases
    dp[0] = nums[0]
    dp[1] = max(nums[0], nums[1])

    for i in range(2, n):
        dp[i] = max(dp[i-2]+nums[i], dp[i-1])

    return dp[-1]

```

After shrinking the space

```

def rob(self, nums):
    if not nums:
        return 0
    n = len(nums)
    if n == 1:
        return nums[0]
    if n == 2:
        return max(nums[0], nums[1])

```

```

# basecases
dp_2 = nums[0]
dp_1 = max(nums[0], nums[1])
for i in range(2, n):
    dp_i = max(dp_2+nums[i], dp_1)
    dp_2 = dp_1
    dp_1 = dp_i
return dp_i

```

#213 In this case, the selection becomes a cycle, then the figure shown below shows the situations. we only compare situation 2 and 3 since all of the elements are non-negative, situation 1 will therefore no need to compare with the other because it has fewer elements.

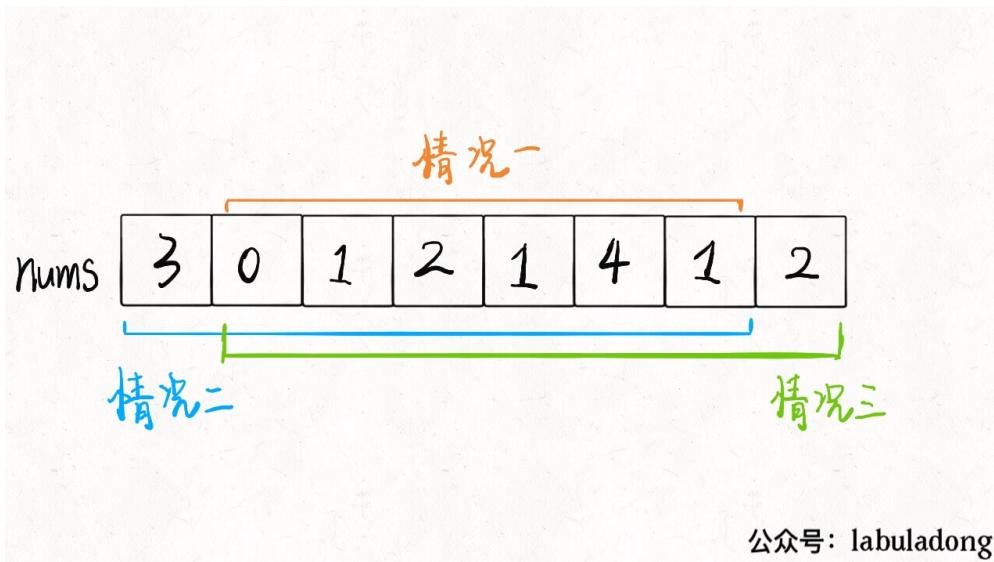


Figure 5.31: Robber 3

```

def rob(self, nums):
    if not nums:
        return 0
    n = len(nums)
    if n == 1:
        return nums[0]
    if n == 2:
        return max(nums[0], nums[1])

    return max(self.helper(nums[:-1]), self.helper(nums[1:]))

def helper(self, nums):

```

```

n = len(nums)
if n == 2:
    return max(nums[0], nums[1])
dp_2 = nums[0]
dp_1 = max(nums[0], nums[1])
dp_i = 0
for i in range(2, n):
    dp_i = max(dp_2+nums[i], dp_1)
    dp_2 = dp_1
    dp_1 = dp_i
# print(dp_i)
return dp_i

# 337

def rob(self, root):
    #Compare grandparent + max of grandchildren(l.l + l.r + r.l + r.r) vs max of children
    def dfs(root):
        if not root:
            return 0,0,0

        l,ll,lr = dfs(root.left)
        r,rl,rr = dfs(root.right)

        return max(root.val + ll + lr + rl + rr, l + r), l, r
    return dfs(root)[0]

```

5.7.15 RegExp Match

Given an input string (s) and a pattern (p), implement regular expression matching with support for '.' and '*'.

- '.' Matches any single character.
- '*' Matches zero or more of the preceding element.

此外，这个问题也可以用动态规划的思路来解决。我们定义状态 $P[i][j]$ 为子串 $s[0, i)$ 和 $p[0, j)$ 是否匹配，能匹配为真，反之为假，然后状态转移方程则可以分为以下三种情况：

1. 如果 $p[j-1] \neq '*' \&& (s[i-1] == p[j-1] \text{ || } p[j-1] == '.')$ ，说明当前两个字符可以匹配且没有遇到 '*'，那么此时 $P[i][j] = P[i-1][j-1]$ ，也即若 $s[0, i-1)$ 和 $p[0, j-1)$ 匹配，则 $s[0, i)$ 和 $p[0, j)$ 也能匹配；
2. 如果 $p[j-1] == '*'$ ，说明当前字符为 '*'，并且我们用其匹配零个字符，那么 $P[i][j] = P[i][j-2]$ ，也即若 $s[0, i)$ 和 $p[0, j-2)$ 匹配，则跳过 $p[j-2], p[j-1]$ 两个元素后 $s[0, i)$ 和 $p[0, j)$ 也能匹配；
3. 如果 $p[j-1] == '*'$ ，说明当前字符为 '*'，并且我们用其匹配至少一个字符，而且还需满足 $s[i-1] == p[j-2] \text{ || } p[j-2] == '.'$ ，那么 $P[i][j] = P[i-1][j]$ ，也即若 $s[0, i-1)$ 和 $p[0, j)$ 匹配，那么我们用这个 '*' 来匹配 $s[i-1]$ 后， $s[0, i)$ 和 $p[0, j)$ 也就能够匹配。至少一次是说这里 '*' 已经被用了一次，而前面 $s[0, i-1)$ 和 $p[0, j)$ 的匹配也有可能会用到。

其中 2 和 3 只需满足一个即可匹配，代码如下。

Figure 5.32: RegExp Transition Functions

```
def isMatch(self, s, p):
    n1 = len(s)
    n2 = len(p)

    if n1 != 0 and n2 == 0:
        return False

    dp = [[False for _ in range(n2+1)] for _ in range(n1 + 1)]
    dp[0][0] = True

    for j in range(2, n2+1):
        if p[j - 1] == '*':
            dp[0][j] = dp[0][j - 2]

    for i in range(1, n1 + 1):
        for j in range(1, n2 + 1):
            if p[j - 1] != '*':
                dp[i][j] = (p[j - 1] == '.' or p[j - 1] == s[i - 1])
                and dp[i - 1][j - 1]
            else:
                dp[i][j] = dp[i][j - 1] or dp[i][j - 2] or ((p[j - 2] == '.'
                    or p[j - 2] == s[i - 1]) and dp[i-1][j])

    return dp[-1][-1]
```

5.7.16 Knuth-Morris-Pratt (KMP)

Original Text: s, length: n1, Pattern sequence: p, length is n2. The KMP algorithm is to find substring p inside s. If not found return -1.

First, to understand what is KMP, it's necessary to figure out what is Partial Match Table (PMT). Look at the table below.

char:	a	b	a	b	a	b	c	a
index:	0	1	2	3	4	5	6	7
value:	0	0	1	2	3	4	0	1

Figure 5.33: PMT

The value inside is the PMT, how is it calculated? The first is to know prefix and suffix. For a string 'aba', the prefix set is 'a', 'ab'. And the suffix set is 'ba', 'a'. The intersection is a. Then the value is 1. So in the table above. The value is calculated like the intersection.

Having known the meaning of this table, how do we boost the search of finding a certain string?

5.8 Two Pointers

5.9 HashTable

#1 TwoSum I, a basic form of these problems, given an array and a integer **target**. Return the index of these 2 numbers.

A simple brute force solution:

```
def twoSum(self, nums, target):
    """
    :type nums: List[int]
    :type target: int
    :rtype: List[int]
```

```

    """
    l = len(nums)
    for i in range(l):
        for j in range(i+1, l):
            if nums[j] == target - nums[i]:
                return [i, j]

    return [-1, -1]

```

We could use hashtable here, since the search complexity of hashtable is $O(n)$, then it'll be easy to find the number and return the index. And the key in the hashtable is the element $\text{nums}[i]$ and value is the index i . It is better to use `collections.defaultdict()`, which will be faster.

```

def twoSum(self, nums, target):
    hashtable = {}
    for i in range(len(nums)):
        hashtable[nums[i]] = i

    for i in range(len(nums)):
        if target - nums[i] in hashtable and hashtable[target - nums[i]] != i:
            return [i, hashtable[target - nums[i]]]

    return []

```

The problem could also be done by using two pointers technique, the prerequisite is that the the given array is sorted.

Chapter 6

Tensorflow & PyTorch

Hello, here is some text without a meaning. This text should show what a printed text will look like at this place. If you read this text, you will get no information. Really? Is there no information? Is there a difference between this text and some nonsense like “Huardest gefburn”? Kjift – not at all! A blind text like this gives you information about the selected font, how the letters are written and an impression of the look. This text should contain all letters of the alphabet and it should be written in of the original language. There is no need for special contents, but the length of words should match the language.

Chapter 7

Spark

7.1 Rationale of using Spark

Engineered from the bottom-up for performance, Spark can be 100x faster than Hadoop for large scale data processing by exploiting in memory computing and other optimizations. Spark is also fast when data is stored on disk, and currently holds the world record for large-scale on-disk sorting. Therefore Spark is suitable for the big data processing.

Spark has easy-to-use APIs for operating on large datasets. This includes a collection of over 100 operators for transforming data and familiar data frame APIs for manipulating semi-structured data.

Spark comes packaged with higher-level libraries, including support for SQL queries, streaming data, machine learning and graph processing. These standard libraries increase developer productivity and can be seamlessly combined to create complex workflows.

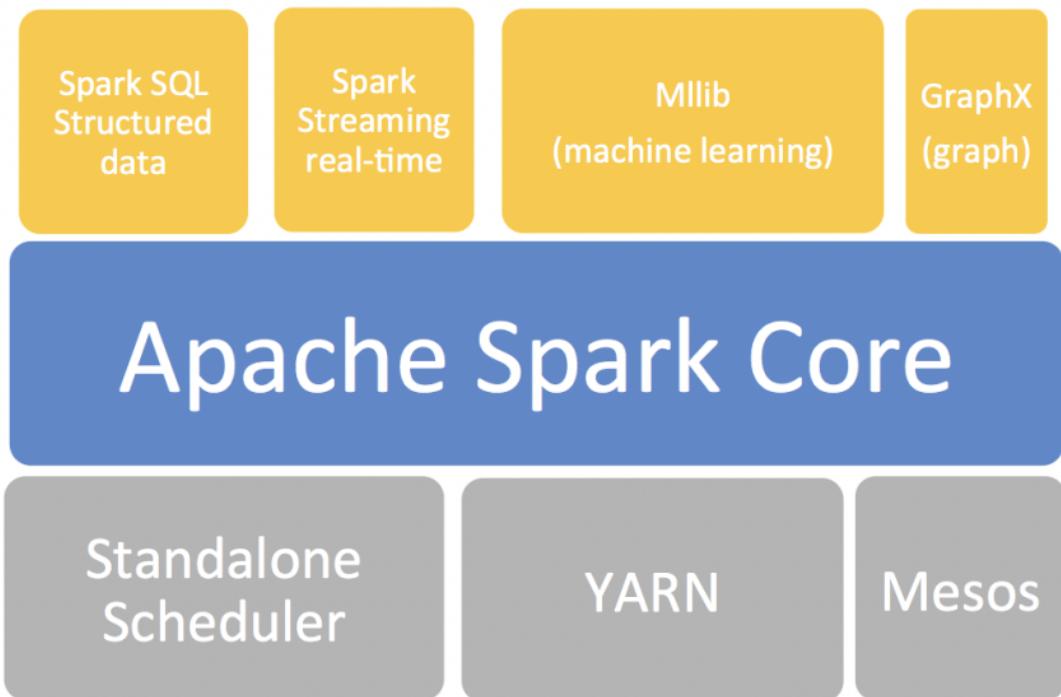


Figure 7.1: Spark Stack

7.2 Core Concept and How Spark Work

- Job: A piece of code which reads some input from HDFS or local, performs some computation on the data and writes some output data.
- Stages: Jobs are divided into stages. Stages are classified as a Map or reduce stages (It's easier to understand if you have worked on Hadoop and want to correlate). Stages are divided based on computational boundaries, all computations (operators) cannot be Updated in a single Stage. It happens over many stages.
- Tasks: Each stage has some tasks, one task per partition. One task is executed on one partition of data on one executor (machine).
- DAG: DAG stands for Directed Acyclic Graph, in the present context its a DAG of operators.
- Executor: The process responsible for executing a task.
- Master: The machine on which the Driver program runs
- Slave: The machine on which the Executor program runs

7.2.1 Spark Component

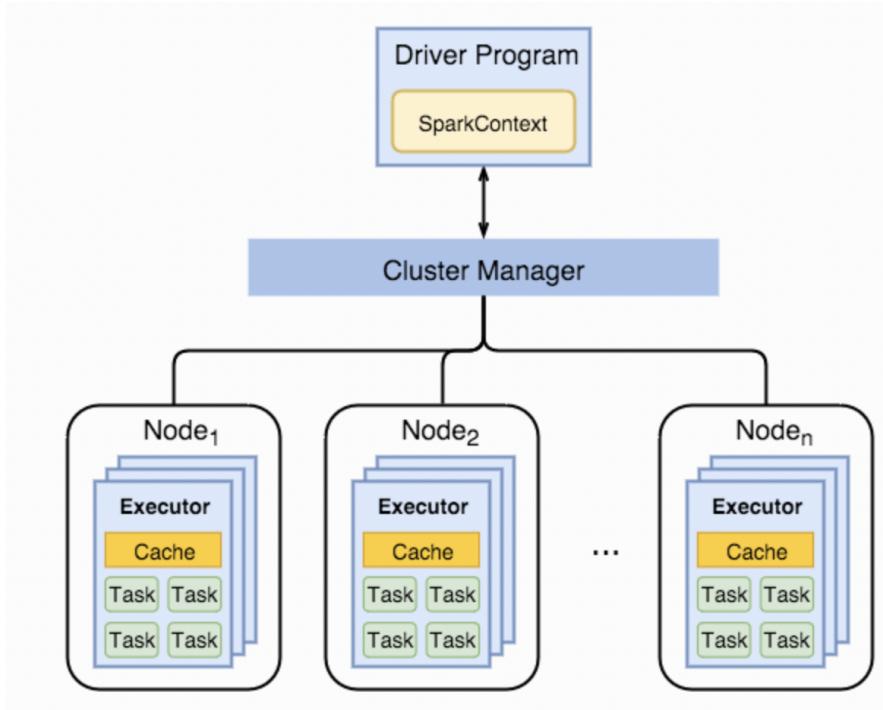


Figure 7.2: Spark Component

- Driver: Separate process to execute user applications, creates SparkContext to schedule jobs execution and negotiate with cluster manager
- Executors: run tasks scheduled by driver, store computation results in memory, on disk or off-heap and interact with storage systems
- Cluster Manager: Mesos, YARN, SparkStandalone

Spark Driver contains more components responsible for translation of user code into actual jobs executed on cluster.

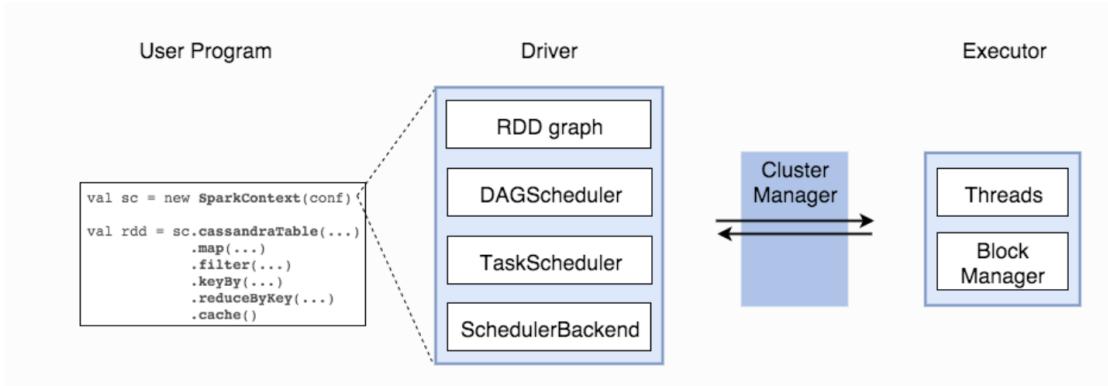


Figure 7.3: spark driver

- **SparkContext**
 - represents the connection to a Spark cluster, and can be used to create RDDs, accumulators and broadcast variables on that cluster
- **DAGScheduler**
 - computes a DAG of stages for each job and submits them to TaskScheduler determines preferred locations for tasks (based on cache status or shuffle files locations) and finds minimum schedule to run the jobs
- **TaskScheduler**
 - responsible for sending tasks to the cluster, running them, retrying if there are failures, and mitigating stragglers
- **SchedulerBackend**
 - backend interface for scheduling systems that allows plugging in different implementations(Mesos, YARN, Standalone, local)
- **BlockManager**
 - provides interfaces for putting and retrieving blocks both locally and remotely into various stores (memory, disk, and off-heap)

7.2.2 Spark Architecture

Spark has a small code base and the system is divided in various layers. Each layer has some responsibilities. The layers are independent of each other.

The first layer is the interpreter, Spark uses a Scala interpreter, with some modifications. As you enter your code in spark console (creating RDD's and applying operators),

Spark creates a operator graph. When the user runs an action (like collect), the Graph is submitted to a DAG Scheduler. The DAG scheduler divides operator graph into (map and reduce) stages. A stage is comprised of tasks based on partitions of the input data. The DAG scheduler pipelines operators together to optimize the graph. For e.g. Many map operators can be scheduled in a single stage. This optimization is key to Sparks performance. The final result of a DAG scheduler is a set of stages. The stages are passed on to the Task Scheduler. The task scheduler launches tasks via cluster manager. (Spark Standalone/Yarn/Mesos). The task scheduler doesn't know about dependencies among stages.

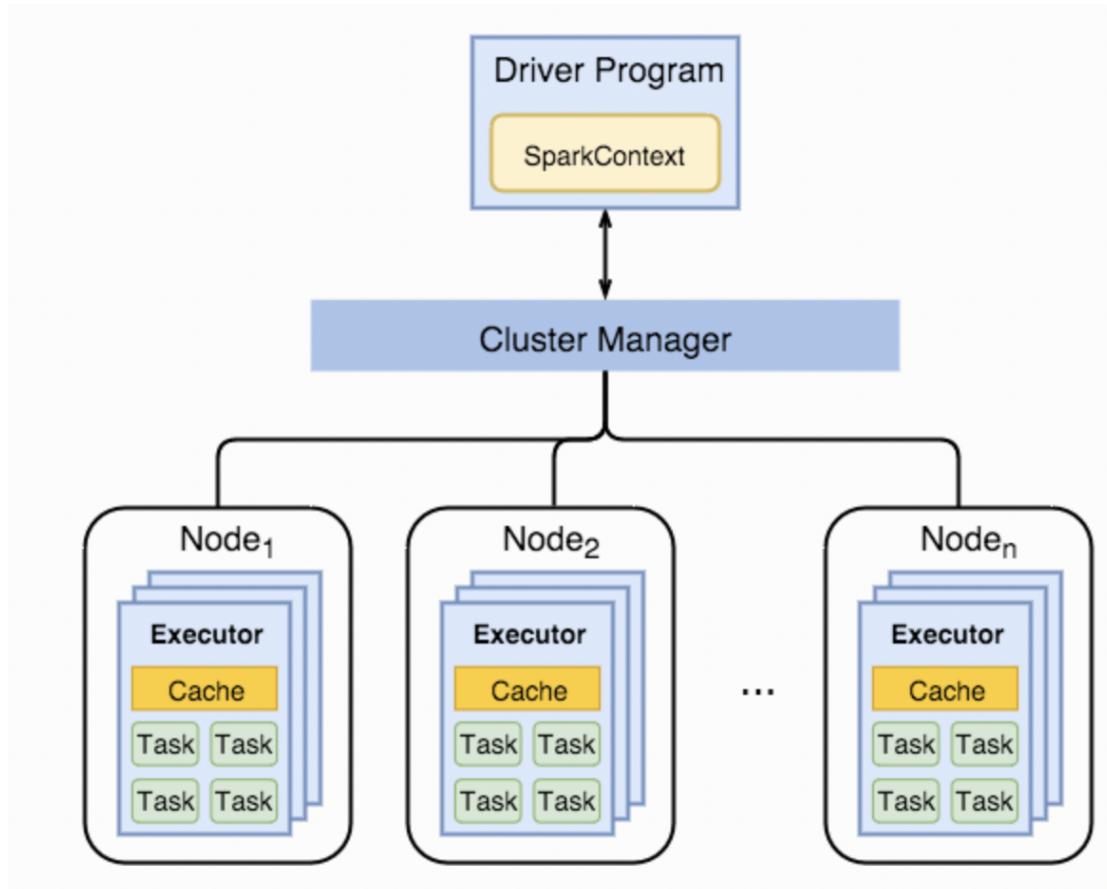


Figure 7.4: spark architecture

7.3 Spark Programming

7.3.1 Difference between SparkSession and SparkContext

Prior to spark 2.0.0, SparkContext was used as a channel to access all spark functionality. The spark driver program uses spark context to connect to the cluster through a resource

manager (YARN orMesos..).

SparkConf is required to create the spark context object, which stores configuration parameter like appName (to identify your spark driver), application, number of core and memory size of executor running on worker node

In order to use APIs of SQL, HIVE , and Streaming, separate contexts need to be created.

```
val conf = new SparkConf()  
  
val sc = new SparkContext(conf)  
  
val hc = new hiveContext(sc)  
  
val ssc = new streamingContext(sc).
```

SPARK 2.0.0 onwards, SparkSession provides a single point of entry to interact with underlying Spark functionality and allows programming Spark with Dataframe and Dataset APIs. All the functionality available with sparkContext are also available in SparkSession.

In order to use APIs of SQL, HIVE, and Streaming, no need to create separate contexts as sparkSession includes all the APIs.

Once the SparkSession is instantiated, we can configure Spark's run-time config properties.

Chapter 8

Docker & Kubernetes

A container is a standard unit of software that packages up code and all its dependencies so the application runs quickly and reliably from one computing environment to another. A Docker container image is a lightweight, standalone, executable package of software that includes everything needed to run an application: code, runtime, system tools, system libraries and settings.

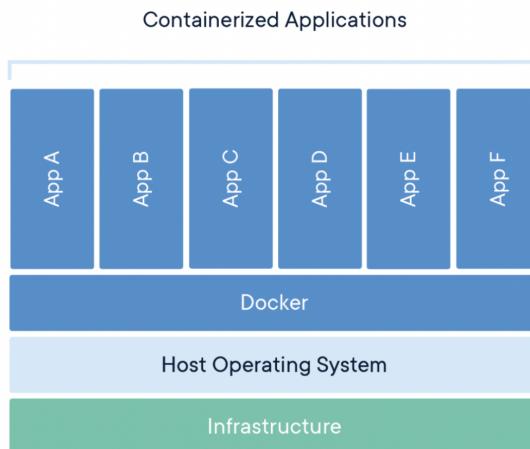


Figure 8.1: Docker in the OS

The operating system support of Virtual machine and Docker container is very different. From the image above, you can see each virtual machine has its guest operating system above the host operating system, which makes virtual machines heavy. While on the other hand, Docker containers share the host operating system, and that is why they are lightweight.

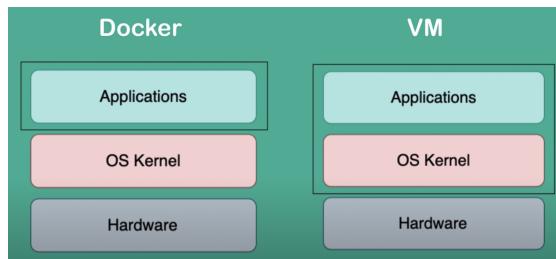


Figure 8.2: Docker v.s. Virtual Machine

8.1 Useful commands

The docker commands could be found in docker doc or docker cheatsheet [pdf]

Most of your images will be created on top of a base image from the Docker Hub registry. Docker Hub contains many pre-built images that you can pull and try without needing to define and configure your own. To download a particular image, or set of images (i.e., a repository), use docker pull

\$ docker pull [OPTIONS] NAME[:TAG|@DIGEST]

List all images that are locally stored with the Docker Engine

\$ docker image ls

List the running containers (add –all to include stopped containers)

\$ docker container ls

Delete an image from the local image store

\$ docker image rm alpine:3.4

The docker run command first creates a writeable container layer over the specified image, and then starts it using the specified command. That is, docker run is equivalent to the API /containers/create then /containers/(id)/start. A stopped container can be restarted with all its previous changes intact using docker start. See docker ps -a to view a list of all containers. You could also use –name option to specify the container name.

\$ docker run [OPTIONS] IMAGE [COMMAND] [ARG...]

An example could be run container in background and print container ID and bind the container's port (see from docker ps -a) and your own OS's port

\$ docker run -p 127.0.0.1:80:8080/tcp -d redis

Show all the running and stopped containers:

```
$ docker ps -a
```

Stop or start a container, to specify a container, you could find the container id from docker ps.

```
$ docker stop [OPTIONS] CONTAINER [CONTAINER...]
$ docker start [OPTIONS] CONTAINER [CONTAINER...]
```

List port mappings or a specific mapping for the container

```
$ docker port CONTAINER [PRIVATE_PORT[/PROTO]]
```

In order to debug our container, there are 2 useful commands we could use, the first one is log, for example, we want to print the last 100 lines of a container's logs

```
$ docker logs --tail 100 web
```

The docker exec command runs a new command in a running container.

```
$ docker exec [OPTIONS] CONTAINER COMMAND [ARG...]
```

An example could be execute commands in container 'web', since this is a lightweight execution bash session, so you won't have all of the command (e.g. curl), but simple ls and cd are supported.

```
$ docker exec -it web /bin/bash
```

8.2 Docker compose

The docker compose file (.yml or .yaml) is a way for us to configure everything we want to run using docker run command, this Compose file is super convenient as we do not have to type all the parameters to pass to the docker run command. We can declaratively do that using a Compose file.

The figure 8.3 shows the mapping from docker run and .yml file. The another good thing about using the compose file is that if you are going to run 2 container in the same network, in the original docker run function, you need to specify -net option and link between this two, but in the compose file it'll create these 2 container within the same network by default.

The command to use is:

```
$ docker compose [-f <arg>...] [options] [COMMAND] [ARGS...]
```

Example:

```
$ docker compose -f docker-compose.yml
```

docker run command	mongo-docker-compose.yaml
<pre>docker run -d \ --name mongodb \ -p 27017:27017 \ -e MONGO-INITDB_ROOT_USERNAME \ =admin \ -e MONGO-INITDB_ROOT_PASSWORD \ =password \ --net mongo-network \</pre>	<pre>version: '3' services: mongodb: image: mongo ports: - 27017:27017 environment: - MONGO..._USERNAME=admin</pre> 

Figure 8.3: docker compose

8.3 Docker workflow & Develop with docker

An good example of how to develop with docker could be found in the doc, The step could be as followed:

- Create a sample Python application
- Create a new Dockerfile which contains instructions required to build a Python image
- Build an image and run the newly built image as a container
- Set up volumes and networking
- Orchestrate containers using Compose
- Use containers for development
- Configure a CI/CD pipeline for your application using GitHub Actions
- Deploy your application to the cloud

The graph 8.4 shows the workflow with docker mainly in the CICD workflow.

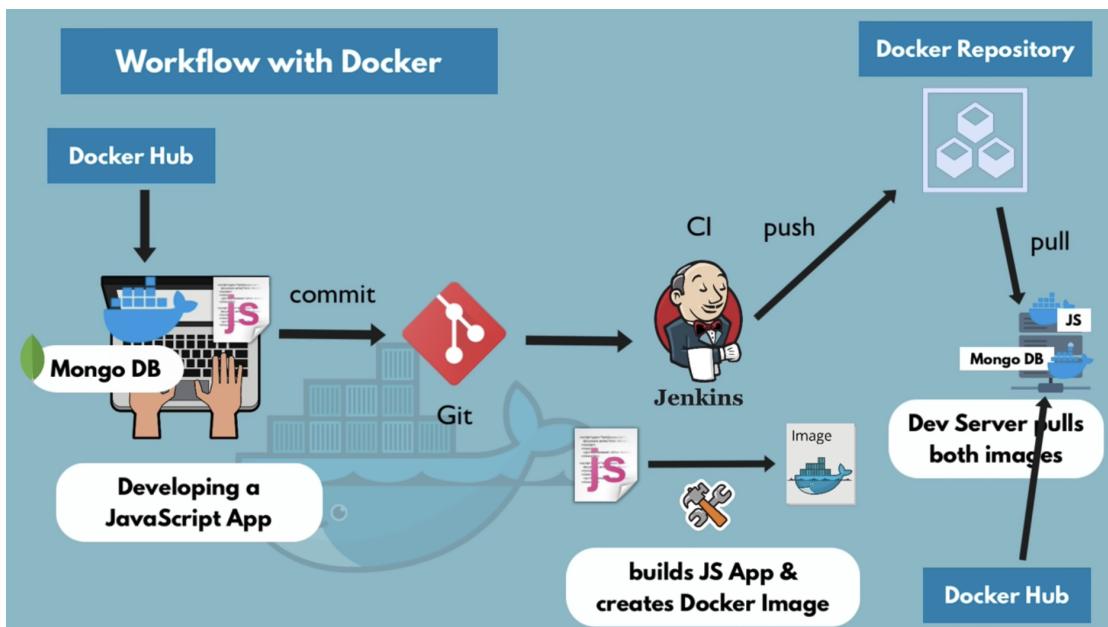


Figure 8.4: Docker CICD work flow

8.4 Dockerfile

Docker can build images automatically by reading the instructions from a Dockerfile. A Dockerfile is a text document that contains all the commands a user could call on the command line to assemble an image. Using **docker build** users can create an automated build that executes several command-line instructions in succession.

Image Environment Blueprint	DOCKERFILE
install node	FROM node
set MONGO_DB_USERNAME=admin	ENV MONGO_DB_USERNAME=admin \
set MONGO_DB_PWD=password	MONGO_DB_PWD=password
create /home/app folder	RUN mkdir -p /home/app
copy current folder files to /home/app	COPY . /home/app
start the app with: "node server.js"	CMD ["node", "server.js"]

Figure 8.5: An dockerfile example

8.5 Private Docker Repository**8.6 Deploy and persistence**