

CSC 600-01 (SECTION 1)  
**Homework 4 - Functional Programming**  
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# CSC 600 HOMEWORK 4 - SCHEME AND FUNCTIONAL PROGRAMMING

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**1. The concept of first class objects is fundamental for Scheme programming. In particular, in Scheme language any function is a first class object. The main properties of a function as a first class object are exemplified by answering the following questions:**

- a) The first class object may be expressed as an anonymous literal value (constant). Show an example of the anonymous function and its use.
- b) The first class object may be stored in variables (i.e. it may have a symbolic name). Show examples of defining and using named functions.
- c) The first class object may be stored in data structures. Show an example of a data structure (e.g. a list) that contains functions.
- d) The first class object may be comparable to other objects for equality. Show an example of comparing functions and lists for equality.
- e) The first class object may be passed as parameter to procedures/functions. Show an example of passing function as an argument to another function.
- f) The first class object may be returned as result from procedures/functions. Show an example of returning a function as a result of another function.
- g) The first class object may be readable and printable. Show examples of:
  - reading function(s) from keyboard,
  - reading function(s) from a file,
  - displaying a function.

The answer is listed on the pages TBD through TBD.

Problem 1a: show an example of the anonymous function and its use:

At the very basic case, an anonymous function is nothing more but a lambda expression - a body of a function. Since it is not associated with any identifier (i.e. when it is unnamed), we may use it only when we explicitly write it inside of other expressions:

```
> ((lambda(x) (* x x)) 2)
4
```

Also, an anonymous function can be used in cases when we would otherwise have to define an inner function (inside another function):

```
(define (add-num-to-each-element-in-list num lst)
  (map (lambda(x) (+ x num)) lst))

> (add-num-to-each-element-in-list 10 '(1 2 3))
(11 12 13)
```

Problem 1b: show examples of defining and using named functions:

Since we already know what a lambda function is, we can now take it and associate it with a name (identifier) - and then we could use just this name to evaluate any expression with it:

```
> (define square (lambda(x) (* x x)))

> square
#<procedure:square>

> (square 2)
4
> (square 4)
16
```

- thus the function definition is a process of binding a lambda expression to some identifier. For simplicity and convenience we could make the same function definition with the use of syntactic sugar and omit 'lambda' keyword. Semantically, such function definition would still remain the same:

```
> (define (square x) (* x x))

> (square 2)
4
> (square 4)
16
```

The reason why we define functions is to follow the Single Responsibility Principle, when each function is doing only one thing. But at certain point we would need to do a function composition, i.e. to define a function that is composed of series of expressions that use previously defined functions:

```
> (define (sum-of-squares lst)
    (apply + (map square lst)))

> (sum-of-squares '(1 2 3 4))
30
```

Problem 1c: show an example of a data structure (e.g. a list) that contains functions:

If we think of a function as an entity that is responsible for only one operation/modification, then chaining different functions to each other to perform sequential modification of the same data can be viewed as a sort of Henry Ford's conveyor belt. In the previous code we considered a case when the number, kinds of functions and their order of evaluation are known and determined, but a data is not. But let's consider a situation when we initially don't know a number, nor kinds of functions, nor their order of evaluation at all. We would need to figure out how to modify a given data dynamically, at runtime. The approach is to pass a list of functions as an extra argument to our master function. With the help of a tail recursion we could take one function at a time from that list, and evaluate it with our data, provided that the arity of these functions match with the number of data arguments. Let's first define a list of functions:

```
> (define list-of-functions '(* - / +))

> list-of-functions
(* - / +)
```

Problem 1d: show an example of comparing functions and lists for equality:

To answer this question we must first understand what equality means, when it comes to comparing functions or lists with each other. For any object to be equal to some other object in Scheme it either needs to have an identical pointer as the other object, or to have the same type and value, or to have the same, equal numerical value. But by definition, the only valid way to compare both functions for their equivalence is to verify that for every valid input both functions produce the same output, and also that both functions have the same domain and range. This means that function equality can really only be correctly defined in terms of operational equivalency; that is, the implementation doesn't matter, only the behavior does. This, of course, is an undecidable problem in any nontrivial language. It is impossible to determine if any two functions are operationally equivalent because, if we could, we could solve the halting problem.

The code listing of main program:

```
#include <stdio.h>
#include <assert.h>
#include "functions.h"

unsigned int maxlen(int *, unsigned int);
```

Auxiliary functions (in separate file "functions.c"):

```
#include <stdio.h>
#include "functions.h"
```

The result of the program execution:

Array a:     1   1   1   2   3   3   5   6   6   6   6   7   9

## 2. Integer plot function (find a smart way to code big integers)

Write a program BigInt(n) that displays an arbitrary positive integer n using big characters of size 7x7, as in the following example for BigInt(170):

```
  @ @      @ @ @ @ @ @ @ @      @ @ @ @ @ @
 @ @ @      @ @      @ @      @ @
```

Write a demo main program that illustrates the work of BigInt(n) and prints the following sequence of big numbers 1, 12, 123, 1234,..., 1234567890, one below the other.

The answer is listed on the pages 7 through 9.

The code listing of the two-dimensional array that stores bit pattern of each BigInt digit. It is declared in the global space (outside of any function).

```
#define NUMBER_OF_ROWS 8

/**
 * Digits are stored as bit patterns of 8-bit unsigned integer (char) numbers.
 *
 * Each digit requires just 8 bytes of storage - which is polynomially smaller
 * than the storage in brute-force approach where each digit is represented by
 * a 2D array of 8x8 characters, with 64 bytes of storage per digit.
 */
```

Main program, excluding the declaration of BIG\_DIGITS array:

```
#include <stdio.h>
#include <math.h>

void BigInt(unsigned int);
unsigned int getNumberOfDigits(unsigned int);

#define NUMBER_OF_BITS 8
#define NUMBER_OF_ROWS 8
```

The result of the program execution:

```
        Blah-blah-blah
asdasd 12 123 12 3123

123123
```

**3. Array processing (elimination of three largest values) (one of many array reduction problems)**

The array  $a(1..n)$  contains arbitrary integers. Write a function  $\text{reduce}(a, n)$  that reduces the array  $a(1..n)$  by eliminating from it all values that are equal to three largest different integers. For example, if  $a=(9, 1, 1, 6, 7, 1, 2, 3, 3, 5, 6, 6, 6, 6, 7, 9)$  then three largest different integers are 6, 7, 9, and after reduction the reduced array would be  $a=(1, 1, 1, 2, 3, 3, 5)$ ,  $n=7$ . The time complexity of the solution should be in  $O(n)$ .

The answer is listed on the pages 11 through 13.

The code listing of the entire program for problem #3:

```
#include <stdio.h>
#include "functions.h"

unsigned int reduce(int *, unsigned int);
void findTop3MaxValuesInArray(int *, unsigned int, int *, int *, int *);
void nullifyTop3MaxValuesInArray(int *, unsigned int, int, int, int);
unsigned int moveZeroesToEndOfArray(int *, unsigned int);
```

Auxiliary functions (in separate file "functions.c"):

```
#include <stdio.h>
#include "functions.h"
```

The result of the program execution:

```
$ gcc -Wall -std=c99 hw2-problem3.c functions.c -O3
$ ./a.out
```



#### 4. Iteration versus recursion (an opportunity for performance measurement)

Make a sorted integer array  $a[i]=i$ ,  $i=0,\dots,n-1$ . Let  $bs(a, n, x)$  be a binary search program that returns the index  $i$  of the array  $a[0..n-1]$  where  $a[i]=x$ . Obviously, the result is  $bs(a, n, x)=x$ , and the binary search function can be tested using the loop

```
for (j=0; j < K; j++)
    for (i=0; i < n; i++)
        if (bs(a, n, i) != i)
            cout << "\nERROR";
```

Select the largest  $n$  your software can support and then  $K$  so that this loop with an iterative version of  $bs$  runs 3 seconds or more. Then measure and compare this run time and the run time of the loop that uses a recursive version of  $bs$ . Compare these run times using maximum compiler optimization (release version) and the slowest version (minimum optimization or the debug version). If you use a laptop, make measurements using AC power, and then the same measurements using only the battery. What conclusions can you derive from these experiments? Who is faster? Why?

The answer is listed on the pages 15 through 19.

The code listing of the entire program for problem #4:

```
#include <stdio.h>
#include <time.h>

#define K 1000                                // system-dependent constant

void initializeArray(int *, int);
int ibs(int *, int, int);
int rbs(int *, int, int, int);
double ibsTest(int *, int);
double rbsTest(int *, int);
```

The result of the program execution with different setup & compiler optimizations:

*# connected to charger, no Wi-Fi, no monitors connected:*

```
$ gcc -Wall -std=c99 hw2-problem4.c -O0
$ ./a.out
Running time of iterative Binary Search: 12.815268 seconds.
```

Initially, with minimal compiler optimization, the implementation of recursive Binary Search

```
$ gcc -Wall -std=c99 hw2-problem4.c -O0
$ ./a.out
Running time of iterative Binary Search: 12.409296 seconds.
```

One possible explanation to that counterintuitive phenomena could be related to behavior defined in operating system. When it activates a power savings mode, it is very likely that certain background processes halt, thus making my program run faster by tens of milliseconds.

## 5. Iteration versus recursion (another opportunity for performance measurement)

Write a recursive function `Frec(n)` that computes Fibonacci numbers. Then write an iterative version of Fibonacci number function `Fit(n)`. Functions `Frec(n)` and `Fit(n)` return the same value but with different performance.

Write the main program that discovers the value `N10` so that `Frec(N10)` runs on your machine exactly 10 seconds. Then measure the run time of `Fit(N10)` and compute how many times is `Fit(N10)` faster than `Frec(N10)`. Show what is `N10` on your machine.

Notes:

1. When you measure the speed, your machine should be disconnected from the Internet, it should use the AC power supply, and it should run only one program (your performance measurement program).
2. In C++ you can measure current time in seconds using the following function:

```
double sec(void)
{
    return double(clock()) / double(CLOCKS_PER_SEC);
}
```

To measure the run time of fast programs you must repeat them many times inside a loop. Take care to eliminate the overhead generated by the loop.

The answer is listed on the pages 20 through 22.

The code listing of the entire program for problem #5:

```
#include <stdio.h>
#include <zconf.h>
#include <time.h>

int Frec(int);
int Fit(int);

double findN(int *);

// using the function pointer to avoid unnecessary code repetition:
double benchmarkFibFunction(int (*f)(int), int);
```

The result of the program execution with different compiler optimizations:

```
$ gcc -std=c99 -Wall hw2-problem5.c -O0
```

```
$ ./a.out && ./a.out
```

Depending on the compiler optimizations, the results may vary. Please **wait...**

As you can see, a slightly more aggressive optimization enables Frec function to find