## Introduction to computational programming Chapter 1 Exercise

Chapter 1 Exercise
Functions and Algorithms
Solutions

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1. Write a function that takes as input a polynomial and returns the polynomial representing the first derivative of the input polynomial. You will be provided two functions, parsePolynomial() and deparsePolynomial to help you in this task. The function parsePolynomial() takes as input a polynomial in x and returns a vector of exponents and a vector of coefficients. The function deparse-Polynomial() takes as input a vector of exponents and a vector of coefficients and returns a polynomial in x.

```
> .calcFirstDerivative <- function(coefs, expons) {
+ new_expons <- expons - 1;
+ new_expons <- new_expons[new_expons != -1]
+ new_coefs <- coefs * expons;
+ new_coefs <- new_coefs[new_expons != -1];
+ new_coefs <- new_coefs[1:length(new_expons)]
+ result <- list(coefs=new_coefs, expons=new_expons);
+ }</pre>
```

<sup>&</sup>lt;sup>1</sup>Helper functions are included at the end of the exercise.

```
> deparsePolynomial <- function(coefficients, exponents) {</pre>
        if (length(exponents) != length(coefficients)) {
            stop ('Exponent vector and coefficient vector must be of
+
        equal length.\n')
        out_string <- '';
       for (ii in seq(along=exponents)) {
  out_string <- paste(out_string, ' + ', as.character());
}</pre>
+
+
        coefficients [ii]),
                                                    'x^', as.character(exponents[ii]),
        sep='');

}
out_string <- gsub('\\+ -', '- ', out_string);
out_string <- gsub('1x', 'x', out_string);
out_string <- gsub('x\\^0', '1', out_string);
out_string <- gsub('x\\^0', '', out_string);
out_string <- gsub('x\\^0', '', out_string);
out_string <- gsub('x\\^1', 'x', out_string);
out_string <- gsub('x\\^1', 'x', out_string);
out_string <- gsub('x\\^1', 'x', out_string);
</pre>
+
       return(out_string);
+ }
```

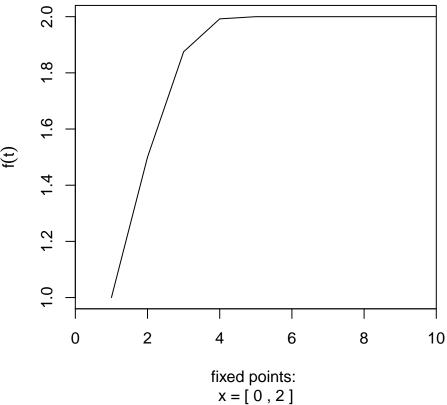
```
> parsePolynomial <- function(inputExpression) {
    numberTerms <- nchar(gsub("[^x]", "", inputExpression));</pre>
    coefficients <- numeric(length=(numberTerms)+1);</pre>
    exponents <- numeric((length=numberTerms)+1);</pre>
    splitTerms <- strsplit(gsub('-', '+-', inputExpression), '</pre>
    \\+')[[1]]
    for (ii in seq(along=splitTerms)) {
      term <- gsub(' ', '', splitTerms[ii]);
if(gsub("[^x]", "", term) == '') {
         term <- paste(term, 'x^0', sep='');
       if (gsub("([\+|-]*)(.*)x.*", "\2", term) == '') { term <- gsub('x', '1x', term);}
       if (gsub(".*x", "", term) == '') {
        term <- paste(term, '^1', sep='');
+
+
       coef <- eval(as.integer(gsub('x.*', '', term)));</pre>
       coefficients[ii] <- coef;</pre>
       exponent <- eval(as.integer(gsub('.*\\^(.*)', '\\1', term
    )));
       exponents[ii] <- exponent;</pre>
+
    o <- -(order(exponents)) + length(exponents) + 1;
    exponents <- exponents[o];</pre>
+
    coefficients <- coefficients[o];</pre>
    result <- list(exponents=exponents, coefficients=
    coefficients)
    return (result);
+ }
```

2. Write a *functor* which takes as an argument a function which is a logistic difference equation. Your functor should find the fixed points of the difference equation and plot its solution as an iterated map.

```
> source('http://rosenbergdm.uchicago.edu/maxima_utilities.R');
> plotImap <- function(f) {
     f_zeros <- mSolve(paste(deparse(body(f)), " = 0", sep=''));
     max_iter <- 10;
     init_value <- f_zeros[1] + diff(f_zeros) / 4;
     x_array <- numeric(length=max_iter);
x_array[1] <- init_value;</pre>
     for (ii in 2: max_iter)
       x_array[ii] <- do.call(f, list(x_array[ii-1]));</pre>
+
+
     fixed_points <- mSolve(paste(deparse(body(f)),</pre>
       " = x", sep=''));
     plot(1:max_iter, x_array, main=paste('Iterated map of f(x))
           \begin{array}{l} \textbf{deparse(} \ body(f) \ ), \ sep=''), \ type='l', \ xaxs='i', \\ xlim=c(0, \ max\_iter), \ xlab='', \ ylab=expression(f(t)), \end{array}
           sub=paste(c("fixed points: \nx = [", fixed_points[1]],
           fixed_points[2], "]"), collapse=" ") );
+
+
     <- function(x)
     x * (2 - x / 2)
```

> plotImap(f)

## Iterated map of f(x)=x \* (2 - x/2)



3. This exercise was not assigned. The solution is included here for you to read if you like. *Quicksort*. Recall exercise 2 from the previous section. There you sorted a list of integers using a method called a *bubble sort*. A more efficient method for sorting is the *quicksort*, given below in pseudocode.

```
function qsort (int array x[n])
   if n == 0 do
        return
   else if n == 1 do
        return x
   else
        int pivot
        int array head
        int array tail
        pivot ← x[0]
        head ← [k] in x with k < pivot
        head ← qsort (head)
        tail ← [k] in x with k >= pivot
        tail ← qsort (tail)
        x ← join(head, pivot, tail)
    return x
```

Pseudocode listing 1 – Quicksort pseudocode.

```
options (width =60);
                          # To make it fit on the page
 qSort <- function (x) {
    if (length(x) < 2) {
      return(x);
    } else {
      pivot \langle x[1];
      x < -x[-1];
      head \leftarrow qSort(x[x \leftarrow pivot]);
      tail <-qSort(x[x >= pivot]);
      return(c(head, pivot, tail));
+
+
> input_vector <- rnorm(n=20) * 100;
> sorted_vector <- qSort(input_vector);
> input_vector
         8.330124
                    23.579529
                                  -2.182591 \quad -250.919532
 [1]
       42.035448 -150.132709
                                -60.302339 214.852173
 [5]
 [9]
     -138.916182
                    45.045513
                                 -47.803216
                                               49.986170
[13]
      136.510902
                    -55.176539
                                  21.170415
                                             -148.596700
[17]
      -53.379635
                    33.240570
                                -34.297054 109.632461
> sorted_vector
 [1]
     -250.919532 \ -150.132709 \ -148.596700 \ -138.916182
                   -55.176539
                                -53.379635
 [5]
      -60.302339
                                              -47.803216
 [9]
       -34.297054
                     -2.182591
                                   8.330124
                                               21.170415
       23.579529
                    33.240570
                                  42.035448
                                               45.045513
[13]
[17]
       49.986170
                   109.632461
                                136.510902
                                              214.852173
```

4. This exercise was not assigned. The solution is included here for you to read if you like. Efficiency. Under ideal conditions, quicksorting a list of n integers involves calling qsort approximately  $n \log n$  times. Under the worst conditions, qsort is called approximately  $n^2$  times. What do these

conditions look like?

Although it may seem counterintuitive, the quicksort is most efficient when its input is "disorganized" and is least efficient when operating on an already sorted array. The command system.time(), which measures how long a computation takes, is used in the following code snippet to demonstrate this difference in efficiency. The options(expression=...) command is necessary to keep R from crashing in the "worst case" scenario.

```
> options(expressions=500000);  # Can cause stack overflow

> inVec <- rnorm(n=1000);

> inVec2 <- qSort(inVec);

> system.time(qSort(inVec));

user system elapsed

0.009  0.000  0.010

> system.time(qSort(inVec2));  # Might crash on some systems

user system elapsed

0.207  0.016  0.284
```

- 5. Using the functor from the previous exercise, plot iterated logistic of the following models and analytically find the fixed points of the models. State whether the solution found computationally approaches the analytically calculated fixed points.
  - (a) f(x) = 2x(1-x)
  - (b) f(x) = 4x(1-x)

