```
In [3]:
    # setup
    from IPython.core.display import display, HTML
    display(HTML('<style>.prompt{width: 0px; min-width: 0px; visibility: collapse}</style>'))
    display(HTML(open('rise.css').read()))

# imports
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns
%matplotlib inline
    sns.set(style="whitegrid", font_scale=1.5, rc={'figure.figsize':(12, 6)})
```

CMPS 2200

Introduction to Algorithms

SPARC

Today's agenda:

- Overview of SPARC language
- Foundation of cost model framework

Why are we learning another language?

- allows us to specify parallel programs concisely
- allows us to analyze runtime of parallel programs
 - particularly for nested recursion
 - recall the recursive fork-join approach to sum an array (lec 2)

SPARC

- based on Standard ML
- functional language

Example SPARC program

let
$$x=2+3$$

$$f(w)=(w*4,w-2)$$

$$(y,z)=f(x-1)$$
 in
$$x+y+z$$
 end

binding: associate entities (data or code) with identifiers.

let expression:

let

 b^+

in

e

end

Expression e is applied using the bindings defined inside **let**.

expression e: describes a computation

• evaluating an expression produces its value

$$x = 2 + 3 = 5$$

 $f(4) \rightarrow (16, 2)$
 $x + y + z = 5 + 16 + 2 = 23$

value: irreducible unit of computation

- e.g.: N, true, -, and
- functions are also values (it is a functional language)

SPARC supports lambda functions like:

lambda x . x+1

lambda (x,y) . x

What do these do?

```
In [11]: f1 = lambda x: x+1 f1(10)
```

Out[11]:

```
In [12]: f2 = lambda x,y : x f2(10,20)
```

```
Out[12]: 10

In [51]: f2(100, 200)

Out[51]: 100
```

Function application

A function application, e_1e_2 , applies the function generated by evaluating e1 to the value generated by evaluating e2.

E.g.,

- if e_1 evaluates to function f(x)
- ullet e_2 evaluates to value v
- ullet apply f to v by substituting v in for x

```
\mathtt{lambda}\left(\left(x,y\right).\ x/y\right)\left(8,2\right)
```

evaluates to 4

Composition

```
sequential composition: (e_1,e_2) parallel composition: (e_1 \mid\mid e_2) e.g. lambda (x,y). (x*x,y*y) vs
```

```
In [44]:
    def compose(g, f):
        """
            Returns a **function** that composes f and g
            """
            return lambda x: g(f(x)) # different from just: g(f(x))

    def meter2cm(d):
        return d * 100

    def cm2inch(d):
        return d / 2.54

# how many inches in a meter?
meter2inch = compose(meter2cm, cm2inch)
meter2inch(1)
```

scoping and recursion

```
x(p)=e vs x={\tt lambda}\ p.\ e When can x be referenced from e? x \text{ is only visible from } e \text{ when defined via the binding } x(p)=e This enables recursive expressions...
```

What does this do?

```
let f(i) = \mathtt{if}\; (i < 2) \; \mathtt{then} \; i \; \mathtt{else} \; i * f(i-1) \mathtt{in} f(5) \mathtt{end}
```

```
In [6]:
    factorial = lambda i: i if i < 2 else i*factorial(i-1)
    factorial(5)</pre>
```

Out[6]: 120

Binary tree

We can also define datatypes recursively like:

```
\texttt{type} \; \mathit{tree} = \mathit{Leaf} \, \mathsf{of} \; \mathbb{Z} \; | \; \mathit{Node} \, \mathsf{of} \; (\mathit{tree}, \mathbb{Z}, \mathit{tree})
```

```
egin{aligned} find \ (t,x) = \ & 	ext{case } t \ & | \ Leaf \ y \Rightarrow x = y \ & | \ Node \ (left,y,right) \Rightarrow \ & 	ext{if } x = y 	ext{ then} \ & 	ext{return true} \ & 	ext{else if } x < y 	ext{ then} \ & 	ext{find } (left,x) \ & 	ext{else} \ & 	ext{find } (right,x) \end{aligned}
```

```
In [7]:  # translated into python...
class Tree:
```

```
def init (self, key, left=None, right=None):
         self.left = left
         self.key = key
         self.right = right
         self.is leaf = left is None and right is None
t = Tree(4,
        Tree (2,
              Tree(1),
              Tree(3)
             ),
        Tree (5,
              Tree (6),
              Tree (7)
         )
def find(t, x):
    print('find t=%d x=%d' % (t.key, x))
    if t.is leaf:
        return t.key == x
    else:
         if x == t.key:
             return True
         elif x < t.key:</pre>
             return find(t.left, x)
         else:
             return find(t.right, x)
find(t, 7)
find t=4 x=7
```

```
find t=4 x=7
find t=5 x=7
find t=7 x=7
Out[7]:
```

Pattern matching

Pattern matching is a way to do typical if .. else statements:

```
egin{aligned} \mathit{find}\,(t,x) = \ & \mathsf{case}\; t \ & |\,\mathit{Leaf}\, y \Rightarrow x = y \ & |\,\mathit{Node}\,(\mathit{left},y,\mathit{right}) \Rightarrow \dots \end{aligned}
```

- Match t against each of the cases.
- When a match is found, evaluate the right hand side of ⇒

What does this do?

```
\mathtt{lambda}\;x\;.\,(\mathtt{lambda}\;y\;.\,f(x,y))
```

Currying

Convert a function of n variables into a sequence of functions with 1 argument each.

- Get specialized functions from more general functions by using composition.
- DRY: no need to repeat function arguments
- Lambda calculus: can define a programming language that only allows functions of one argument
 - easier for proofs!

```
E.g., f(x,y)=x+y^2 lambda x . (\texttt{lambda}\ y . f(x,y))(10)(20) \to lambda y . f(10,y)(20) \to lambda y . (10+y^2)(20) \to 10+20^2 \to 410
```

```
In [9]:
        def curry(f):
            11 11 11
            Given a function f of two variables,
            return a function g that binds the first variable
            and returns a function of the second variable.
             def q(x): # nested function 1
                def h(y): # nested function 2
                    return f(x, y)
                return h # bind x and return function of y
             return g
        def f(x, y):
             return x + y**2
        print(f)
                               # returns f'n g. input: x, output function of y
        print(curry(f))
        print(curry(f)(10)) # returns f'n h. input: y, output f(10,y)
        print(curry(f)(10)(20)) # returns f(10,20)
        print(curry(f)(10)(3)) # returns f(10,3)
```

```
<function f at 0x10444c488>
<function curry.<locals>.g at 0x10444c598>
<function curry.<locals>.g.<locals>.h at 0x10444c730>
410
19
```

Next lecture we will see how SPARC will allow us to analyze the cost of an algorithm.