

HYPERDIMENSIONAL COMPUTING FOR PROTEIN LANGUAGE MODELING

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Michael Fatjanov

Student ID: ...

Supervisor(s): Prof. Dr. Bernard De Baets

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Gent, FILL IN THE DATE

The author,

Michael Fatjanov

The promotor,

Prof. Dr. Bernard De Baets

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SAMENVATTING

nederlandse samenvatting

SUMMARY

insert english summary here...

1. INTRODUCTION

- 1.1 Digital biology, protein sequence research and traditional bioninformatics tools
- 1.2 State-of-the-art, deep learning and protein language modeling
- 1.3 Hyperdimensional computing

BIBLIOGRAPHY

APPENDIX A

MODEL DEVELOPMENT CODING

A.1 Pseudocode of the presented algorithm

```
Algorithm 1: How to write algorithms

Data: this text

Result: how to write algorithm with LTEX2e initialization;

while not at end of this document do

read current;

if understand then

go to next section;

current section becomes this one;

else

go back to the beginning of current section;

end

end
```

A.2 Sensitivity base class code

```
1 import os
2 import numpy as np
3
4 from parameter import *
5 import matplotlib.pyplot as plt
6 from matplotlib.ticker import FixedLocator, MaxNLocator
7
8 class SensitivityAnalysis(object):
9 """
10 Base class for the Sensitivity Analysis
11
12 Parameters
```

```
13
      ParsIn : list
14
           ModPar class instances in list or list of (min, max, 'name')-
15
              tuples
16
17
      Attributes
       -----
18
      ParsIn : list
19
           a list of (min, max, 'name') values,
20
           [(min,max,'name'),(min,max,'name'),...(min,max,'name')]
21
      parmap : dict
22
           tracks the sequence of the parameters
23
24
      Pars : list of ModPar instances
25
           Used when working with the pyFUSE package
      ndim : int
26
27
           number of uncertain input factors
28
      namelist : list
           list of the uncertain input factors used
29
30
       0.000
31
32
      def __init__(self,ParsIn):
33
34
           Check if all uniform distribution => TODO ! if all -> sobol
35
              sampling
           is possible, else, only uniform and normal distribution are
36
              supported
           for using the sobol sampling... Here is still work to do!!
37
38
39
           if isinstance(ParsIn, dict): #bridge with pyFUSE!
40
               dictlist = []
41
               for key, value in ParsIn.iteritems():
42
                   dictlist.append(value)
43
               ParsIn = dictlist
44
               print ParsIn
45
46
           #control for other
47
           self.ParsIn = ParsIn
48
           self.parmap={} #dictionary linking ID and name, since dict
49
              instance has no intrinsic sequence
50
           for i in range(len(ParsIn)):
               if isinstance(ParsIn[i], ModPar): #or isinstance(ParsIn[i],
51
                  pyFUSE.parameter.ModPar):
                   cname = ParsIn[i].name
52
                   self.Pars = ParsIn
53
                   self.ParsIn[i] = (ParsIn[i].min, ParsIn[i].max, cname)
54
```

```
self.parmap[i] = cname
55
56
               elif isinstance(ParsIn[i],tuple):
57
                   if ParsIn[i][0] > ParsIn[i][1]:
58
                       raise Exception('Min value larger than max value')
59
                   if not isinstance(ParsIn[i][0],float) and isinstance(
60
                      ParsIn[i][1],float):
                       raise Exception('Min and Max value need to be float'
61
                           )
                   if not isinstance(ParsIn[i][2],str):
62
                       raise Exception('Name of par needs to be string')
63
                   self.parmap[i] = ParsIn[i][2]
64
                   #create modpar instance of the tuple
65
                   self.Pars=[]
66
                   for par in ParsIn:
67
                       self.Pars.append(ModPar(par[2],par[0],par[1],(par
68
                           [0]+par[1])/2.,'randomUniform'))
               else:
69
                   raise Exception('The input type for sampling not correct
70
                   choose ModPar instance or list of (min,max)-tuples')
71
72
           self.ndim=len(ParsIn)
73
74
           self.namelist = []
75
           for i in range(self.ndim):
76
77
               self.namelist.append(self.parmap[i])
78
      def WritePre(self,filename = 'inputparameterfile', *args, **kwargs):
79
80
           Parameterinputfile for external model, parameters in the columns
81
               files
           and every line the input parameters
82
83
           Parameters
84
           _____
85
           filename : str
86
               name of the textfile to save
87
           *args, **kwargs : args
88
               arguments passed to the numpy savetext-function
89
90
91
           np.savetxt(filename, self.parset2run, *args, **kwargs)
92
           print 'file saved in directory %s'%os.getcwd()
93
94
      def ReadRuns(self,filename, *args, **kwargs):
95
96
```

```
Read model outputs (TODO: do sobol for multiple outputs,
97
               iterating the
98
           post)
           Format is: every output of the ithe MC on ith line
99
100
           output2evaluate can also be made on a other way
101
102
           Parameters
103
            _____
104
           filename : str
105
                name of the textfile to load
106
           *args, **kwargs : args
107
108
                arguments passed to the numpy loadtext-function
109
110
           self.output2evaluate = np.loadtxt(filename, *args, **kwargs)
111
```

A.3 Model input file for PyFUSE model

```
2 ##
       Model Parameter input file
3 ##
       The parameter is defined by his distribution,
4 ##
       boundaries and extra info needed by distribution
5 ##
       provide on each line one parameter with
6 ##
       following information:
  ##
       name : string
7
8 ##
           Name of the parameter
9
  ##
       minval : float
10 ##
           Minimum value of the parameter distribution
11 ##
       maxval : float
           Maximum value of the parameter distribution
12 ##
13 ##
       optquess : float
           Optimal guess of the parameter, must be
14 ##
15 ##
           between min and max value
16 ##
       pardistribution : string
           choose a distributionfrom: randomUniform,
17 ##
           randomTriangular, randomTrapezoidal,
18 ##
           randomNormal, randomLogNormal
19 ##
       *kargs
20 ##
           Extra arguments necessary for the
21 ##
22 ##
           chosen distribution
     Lines with ## marks are neglected
23 ##
25 ## NAME MIN MAX OPTGUESS DISTRIBUTION ARGS*
26 S1max 50. 5000.000 400. randomTriangular 1000.
```

ko 0.01 0.99 0.15 randomUniform timeo 2. 48 24 randomUniform timei 2. 250. 20 randomUniform

56 timeb 200. 10000. 2100. randomUniform

27 S2max 100. 10000.000 1000. randomNormal 500. 25. 28 fitens 0.01 1.0 0.99 randomLogNormal 0.5 0.2 29 firchr 0.050 0.950 0.5 randomTrapezoidal 0.4 0.6 30 fibase 0.050 0.950 0.5 randomUniform 31 r1 0.050 0.950 0.5 randomUniform 32 ku 0.01 1000. 0.044 randomUniform 33 c 0.99 20.0 1. randomUniform 34 alfa 1.000 250. 150. randomUniform 35 psi 1.000 5.0 2.5 randomUniform 36 kappa 0.050 0.950 0.5 randomUniform 37 ki 0.001 1000. 0.00833 randomUniform 38 ks 0.001 10000. 0.5 randomUniform 39 n 1.000 10. 3. randomUniform 40 v 0.00001 0.250 0.004 randomUniform 41 vA 0.001 0.250 0.0015 randomUniform 42 vB 0.001 0.250 0.0015 randomUniform 43 Acmax 0.050 0.950 0.5 randomUniform 44 b 0.001 3.0 0.2 randomUniform 45 loglambda 5.000 10.0 7.5 randomUniform 46 chi 2.000 5.0 3.5 randomUniform 47 mut 0.010 5.0 0.6 randomUniform 48 be 0.99 4. 3.1 randomUniform 49 alfah 0.01 0.99 0.5 randomUniform 50 tg 0.0 0.7 0.3 randomUniform 51 tif 0.0 0.7 0.26 randomUniform 52 tof 0.0 0.7 0.12 randomUniform