## AppendixVIII

March 27, 2020

```
[]: import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     from model import Fuzzification, InferenceEngine
     import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     plt.rcParams['font.weight'] = 'bold'
     plt.rcParams['font.size'] = 12
     plt.rcParams['axes.labelweight'] = 'bold'
     plt.rcParams['lines.linewidth'] = 2
     plt.rcParams['axes.titleweight'] = 'bold'
     class Fuzzification:
         dx = 0.001 \# dx for linguistic variable function
         nd = 3
         #N = 1000
         Ostaticmethod
         def norm_type(iv,typeofnorm,discourse):
             if typeofnorm == 'lower_is_better':
                 x = Fuzzification.trapezoid(z= discourse,
                                             c_l=iv.min(),
                                             tc=iv.min(),
                                             Tc=iv.min(),
                                             c_u=iv.max());
             elif typeofnorm == 'middle_is_better':
                 x = Fuzzification.trapezoid(z= discourse,
                                             c_l=iv.min(),
                                             tc=iv.mean(),
                                             Tc=iv.mean(),
                                             c_u=iv.max());
             elif typeofnorm == 'higher_is_better':
```

```
x = Fuzzification.trapezoid(z=discourse,
                                          c_l=iv.min(),
                                          tc=iv.max(),
                                          Tc=iv.max(),
                                          c_u=iv.max())
       else:
           raise 'Error'
       return x
   Oclassmethod
   def trapezoid(cls,z,c_l,tc,Tc,c_u):
        Trapezoid function that can be used to create linguistic values or \Box
\hookrightarrow normalization curves.
       Rule #1 of Fuzzy System: Completeness of Inputs - Make sure z covers_{\sqcup}
\hookrightarrow the full universe of discourse.
          assert c_l<=tc, 'input values dont make sense for trapezoid'
         assert tc<=Tc, 'input values dont make sense for trapezoid'
         assert Tc<=c_u, 'input values dont make sense for trapezoid'
       if (c_l==tc) and (c_l==Tc): # DECREASING LINE
            #print('decreasing line')
            out = np.piecewise(z,
                                 [z<c_1, z>c_u, (z>=Tc)&(z<=c_u)],
                                 [0 , 0 , lambda z: -z/(c_u-c_1) + c_u/
\hookrightarrow (c_u-c_1)])
       elif (c_u==tc) and (c_u==Tc): # INCREASING LINE
            #print('increasing line')
            out = np.piecewise(z,
                                 [z<c_1, z>c_u, (z>=c_1)&(z<=tc)],
                                 [0 , 0 , lambda z: z/(c_u-c_1) - c_1/
\hookrightarrow (c_u-c_1)])
       else: # TRAPEZOID AND TRIANGLE
            #print('trapezoid')
            out = np.piecewise(z,
                                 [z <= c_1, z >= c_u, (z > c_1) & (z < tc)
                                                                             , ⊔
\hookrightarrow (z>=tc)&(z<=Tc), (z>Tc)&(z<c_u)],
                                              , lambda z: (z-c_1)/(tc-c_1), 1
                                 Γ0
                                    , 0
         , lambda z: (c_u-z)/(c_u-Tc)])
       out = out.round(cls.nd) #getrid of annoying decimals
       return out
```

```
@classmethod
   def wms(cls):
       Linguistic Variable WMS
       11 11 11
       x= np.arange(0,1+cls.dx,cls.dx).round(cls.nd)
       medium_val = 0.7
       df = pd.DataFrame(data = {'w':Fuzzification.
\rightarrowtrapezoid(x,0,0,0,medium_val),
                                  'm':Fuzzification.
→trapezoid(x,0,medium_val,medium_val,1),
                                  's':Fuzzification.
→trapezoid(x,medium_val,1,1,1)},
                          index = x)
       Fuzzification.check_ruspini_partition(df)
       Fuzzification.check_consistency(df)
       return df.round(cls.nd)
   @classmethod
   def vbbagvg(cls):
       Linguistic Variable VBBAGVG
       x=np.arange(0,1+cls.dx,cls.dx).round(cls.nd)
       d = 0.25
       data = {}
       for i,l in enumerate(['vb','b','a','g','vg']):
           data[1] = Fuzzification.trapezoid(x,d*(i-1),d*(i),d*(i),d*(i+1))
       df = pd.DataFrame(data = data,index = x)
       cls.check ruspini partition(df)
       cls.check_consistency(df)
       return df.round(cls.nd)
   @classmethod
   def elvllflifhhvheh(cls):
       Linquistic Variable elvllflifhhvheh
       x=np.arange(0,1+cls.dx,cls.dx).round(cls.nd)
       d = 0.125
       data = \{\}
       for i,l in enumerate(['el','vl','l','fl','i','fh','h','vh','eh']):
           data[1] = Fuzzification.trapezoid(x,d*(i-1),d*(i),d*(i),d*(i+1))
       df = pd.DataFrame(data = data,index = x)
       cls.check_ruspini_partition(df)
       cls.check_consistency(df)
```

```
return df.round(cls.nd)
   Ostaticmethod
   def check_ruspini_partition(df):
       Special case of Rule #2 of Fuzzy System: Consistency of Unions for WMS_{\sqcup}
\hookrightarrow and VBBAGVG
       assert any(df.sum(axis='columns').apply(lambda x: round(x,2)==1)) ==__
→True , 'Ruspini Partition Not Satisfied'
   Ostaticmethod
   def check_consistency(df):
       11 11 11
       Rule #2 of Fuzzy System: Consistency of Unions - for any input, the ⊔
\hookrightarrowmembership functions of all the fuzzy sets it belongs too should be less\sqcup
\hookrightarrow than or equal to 1.
       assert any(df.sum(axis='columns').apply(lambda x: round(x,2)<=1)) ==__
\hookrightarrowTrue , 'Not Consistent'
   Ostaticmethod
   def primary(primary_indicator,basic_indicators,indicator_type):
       #construct linguistic value lookuptables
       wms = Fuzzification.wms()
       #vbbagvg = Fuzzification.vbbagvg()
       #primary indicator db (the right excel tab)
       pi_db = pd.read_excel('./databases/indicator_db.xlsx',_
⇒sheet_name=primary_indicator).round(Fuzzification.nd)
       # CONSTRUCT NORMALIZATION CURVES
       curves = []
       for basic_indicator in basic_indicators.keys():
            assert pi_db[pi_db[indicator_type] == basic_indicator].shape[1] == 9_u
→, 'Warning, not expected shape' #check, sensitve to number of basic_
\rightarrow indicators, CHANGE ME
            basic_df = pi_db[pi_db[indicator_type] == basic_indicator]
            #print(basic_indicator)
            iv = basic_df['intensive_value']
            discourse = np.arange(iv.min(),iv.max()+Fuzzification.
\rightarrowdx,Fuzzification.dx).round(Fuzzification.nd)
           x = Fuzzification.norm_type(iv=iv,
```

```
→ typeofnorm=basic_indicators[basic_indicator],
                                     discourse=discourse)
          ncurve = pd.Series(data=x.round(Fuzzification.nd),
                              index=discourse.round(Fuzzification.nd),
                              name=basic indicator)
          curves.append(ncurve)
       # PLOT/SAVE NORMALIZATION CURVES
      for curve in curves:
          fig = plt.figure(figsize=(10,5));plt.title('Normalization Curve:
→{}'.format(curve.name));
          curve.plot();
          plt.grid();plt.xlabel('z [{}]'.
→plt.ylabel('x'); #plt.legend(basic_indicators.keys());
          fig.savefig('./outputs/normalization_curves/{}.png'.format(curve.
→name), dpi=300, bbox_inches='tight')
       #FUZZIFICATION
      frames = []
      for curve in curves:
          df_base = pi_db[pi_db[indicator_type] == curve.name].

→drop(['raw_value', 'raw_value_units', 'intensive_units', 'source'], axis='columns').
→reset_index(drop=True) #reset index so that it concats properly
          z = df base['intensive value'].values
          x = curve.loc[z] # pass through normalization curve
          wms_v = wms.loc[x.values].reset_index(); #reset index to make_
\rightarrow concat work
          df_out = pd.concat([df_base,wms_v],axis='columns')
          assert df_base.shape[0] == df_out.shape[0], 'different amount of_
⇔rows, error¹
          frames.append(df_out)
      fuzz = pd.concat(frames).reset index(drop=True)
       #INFERENCE
      frames2=[]
      for company in fuzz['company'].unique():
          frames = []
          for year in fuzz['year'].unique():
              company_year = fuzz[(fuzz['company'] == company) &
                                  (fuzz['year'] == year)]
              #print(year, company)
```

```
b_indicators =__
 → [company_year[company_year[indicator_type] == indicator].iloc[0] for indicator_
 →in basic_indicators]
                frames.append(InferenceEngine.b_s(b_indicators)) #apply_
 → inference engine
            frames2.append(pd.concat(frames,axis='columns').T)
        secondary = pd.concat(frames2).reset_index(drop=True)
        return secondary
class InferenceEngine:
    dx = 0.01 \# dx \text{ for linguistic variable function}
    nd = 2
    Ostaticmethod
    def osus():
        return None
    Ostaticmethod
    def defuzzification():
        return None
    #primary indicators
    Oclassmethod
    def s_p(cls,secondary_indicators,primary_ind_name,indictor_type):
        SECONDARY TO PRIMARY INFERENCE ENGINE &
        PRIMARY TO OSUS
        *** ONLY FOR 2 secondary INDICATOR INPUTS
        HHHH
        #checks
        assert all(i['company'] == secondary_indicators[0]['company'] for i in__
 ⇒secondary_indicators), 'secondary indicators included dont have the same u
→company¹
        secondary_ind_company = secondary_indicators[0]['company']
        assert all(i['year'] == secondary_indicators[0]['year'] for i in_
⇒secondary_indicators), 'secondary indicators included dont have the same u
 ⇔year'
        secondary_ind_year = secondary_indicators[0]['year']
        #search the rule base for the secondary indicator
```

```
rb = pd.read_excel('./databases/rulebase.xlsx',__
⇒sheet_name=primary_ind_name,skiprows=13)
       lval = []
       indices = []
       if len(secondary indicators) == 2: #if there are two secondary indicators
            for in1_lv in ['vb','b','a','g','vg']:
                for in2_lv in ['vb','b','a','g','vg']:
                    lval.append(secondary_indicators[0][in1_lv] *__
→secondary_indicators[1][in2_lv]) #LARSEN IMPLICATIONsensitive to the number_
\hookrightarrow of indicators
                    rule = rb[(rb[secondary indicators[0][indictor type]] ==___
\rightarrowin1_lv) &
                               (rb[secondary_indicators[1][indictor_type]] ==__
→in2 lv)] #selecting the right row from the rulebase, sensitive to the number
\hookrightarrow of indicators
                    indices.append(rule[primary_ind_name].iloc[0]) #
       elif len(secondary_indicators) == 3: #if there are three secondary_
\rightarrow indicators
           for in1_lv in ['vb','b','a','g','vg']:
                for in2_lv in ['vb','b','a','g','vg']:
                    for in3_lv in ['vb','b','a','g','vg']:
                        lval.append(secondary_indicators[0][in1_lv] *__
⇒secondary_indicators[1][in2_lv] * secondary_indicators[2][in3_lv]) #LARSEN_⊔
\hookrightarrow IMPLICATIONs ensitive to the number of indicators
                        rule = rb[(rb[secondary_indicators[0][indictor_type]]__
→== in1 lv) &
                                   (rb[secondary_indicators[1][indictor_type]]
\rightarrow == in2_lv) &
                                   (rb[secondary_indicators[2][indictor_type]]_
→== in3 lv)] #selecting the right row from the rulebase, sensitive to the
\rightarrow number of indicators
                        indices.append(rule[primary_ind_name].iloc[0]) #
       else:
           print('WARNING, THER ARE SOME OTHER NUMBER OF secondary INDICATORS')
       s = pd.Series(data=lval,
                      index=indices).round(cls.nd)
       s = s.groupby(s.index).sum() #add up all similar vals
       s['primary'] = primary_ind_name
       s['year'] = secondary_ind_year
       s['company'] = secondary ind company
       return s
```

```
#basic indicators to secondary indicators
   @classmethod
   def b_s(cls,basic_indicators):
       BASIC TO SECONDARY INFERENCE ENGINE
       *** ONLY FOR 2 BASIC INDICATOR INPUTS
       11 11 11
       #checks
       assert all(i['company'] == basic_indicators[0]['company'] for i in_
⇒basic_indicators), 'basic indicators included dont have the same company'
       secondary_ind_company = basic_indicators[0]['company']
       assert all(i['secondary'] == basic_indicators[0]['secondary'] for i in__
⇒basic indicators), 'basic indicators included dont have the same secondary |
→indicator'
       secondary_ind_name = basic_indicators[0]['secondary'] #status
       assert all(i['year'] == basic_indicators[0]['year'] for i in_
⇒basic_indicators), 'basic indicators included dont have the same year'
       secondary_ind_year = basic_indicators[0]['year']
       #search the rule base for the secondary indicator
       rb = pd.read_excel('./databases/rulebase.xlsx',__
→sheet_name=secondary_ind_name,skiprows=13)
       lval = []
       indices = []
       if len(basic indicators) == 2: #if there are two basic indicators
           for in1_lv in ['w','m','s']:
               for in2_lv in ['w','m','s']:
                   lval.append(basic_indicators[0][in1_lv] *__
→basic_indicators[1][in2_lv]) #LARSEN IMPLICATIONsensitive to the number of
\rightarrow indicators
                   rule = rb[(rb[basic_indicators[0]['basic']] == in1_lv) &
                              (rb[basic_indicators[1]['basic']] == in2_lv)]__
→#selecting the right row from the rulebase, sensitive to the number of
\rightarrow indicators
                   indices.append(rule[secondary ind name].iloc[0]) #
       elif len(basic_indicators)==3: #if there are three basic indicators
           for in1_lv in ['w','m','s']:
               for in2 lv in ['w','m','s']:
                   for in3_lv in ['w','m','s']:
```

```
lval.append(basic_indicators[0][in1_lv] *__
 ⇒basic_indicators[1][in2_lv] * basic_indicators[2][in3_lv]) #LARSEN_
 → IMPLICATIONsensitive to the number of indicators
                       rule = rb[(rb[basic indicators[0]['basic']] == in1 lv) &
                                  (rb[basic_indicators[1]['basic']] == in2_lv) &
                                  (rb[basic indicators[2]['basic']] == in3 lv)]___
 →#selecting the right row from the rulebase, sensitive to the number of
 \rightarrow indicators
                       indices.append(rule[secondary_ind_name].iloc[0]) #
        else:
           print('WARNING, THER ARE SOME OTHER NUMBER OF BASIC INDICATORS')
       s = pd.Series(data=lval,
                      index=indices).round(cls.nd)
       s = s.groupby(s.index).sum() #add up all similar vals
       s['secondary'] = secondary_ind_name
       s['year'] = secondary_ind_year
       s['company'] = secondary_ind_company
       return s[['secondary','company','year','vb','b','a','g','vg']]
struct = {'wealth':{
                    'status':{'roe':'middle_is_better',
                              'roa': 'higher is better'},
                    'pressure':{'CF_CAPEX':'higher_is_better',
                                'dividend payout ratio': 'middle is better'},
                    'response':{'debt to equity ratio':'middle_is_better',
                                'operating expenses':'lower is better',
                                'effective tax rate':'lower_is_better'}
                  },
          'ecos':{
                    'air':{'clean generation':'lower_is_better', #bc we want to_
⇒see a low percentage of coal
                           'CO2 emissions':'lower is better'},
                    'land':{'spills':'lower_is_better',
                            'solid waste':'lower is better',
                            'hazardous waste':'lower_is_better'},
                    'water':{'total water withdrawal':'lower is better',
                             'percent water consumed':'lower_is_better'}
                  },
          'hums':{
                    'health':{'fatalities':'lower_is_better', #bc we want to_
 ⇒see a low percentage of coal
```

```
'osha recordable rate (ticr)':'lower_is_better'},
                  'polic':{'lobbying spending':'lower_is_better',
                         'charitable giving':'higher_is_better'}
                 }
        }
#_____
print('Starting Basic -> Secondary')
frames2 = \Pi
for primary_indicator in struct.keys():
     if (primary_indicator == 'wealth') or (primary_indicator == 'ecos'):
         continue # skip for now
   frames = []
   for secondary_indicator in struct[primary_indicator]:
       print('{}-{}'.format(primary_indicator ,secondary_indicator))
       frames.append(Fuzzification.primary(primary_indicator=primary_indicator,
→basic_indicators=struct[primary_indicator][secondary_indicator],
                                       indicator type='basic'))
   secondary_agg = pd.concat(frames,axis='index').reset_index(drop=True)
   secondary_agg.to_excel('./outputs/{}_secondary_agg.xlsx'.
→format(primary_indicator))
   frames2.append(secondary_agg)
print('Starting Secondary -> Primary')
frames3=[]
for primary_ind_name in struct.keys():
   print(primary_ind_name)
   secondary_agg = pd.read_excel('./outputs/{}_secondary_agg.xlsx'.
→format(primary_ind_name))
   indicator_type = 'secondary'
   #TNFF.R.F.NCF.
   frames2=[]
   for company in secondary_agg['company'].unique():
       frames = []
       for year in secondary_agg['year'].unique():
```

```
company_year = secondary_agg[(secondary_agg['company'] == company) &
                                    (secondary_agg['year'] == year)]
          #print(year, company)
          s_indicators =_
→ [company_year[company_year[indicator_type] == indicator].iloc[0] for indicator_
→in secondary_agg[indicator_type].unique()]
          frames.append(InferenceEngine.
→s_p(s_indicators,primary_ind_name,indicator_type)) #apply inference engine
       frames2.append(pd.concat(frames,axis='columns').T)
   primary = pd.concat(frames2).reset_index(drop=True)
   frames3.append(primary)
primary_agg = pd.concat(frames3)
primary_agg.to_excel('./outputs/primary_agg.xlsx')
print('Starting Primary -> OSUS')
primary_agg = pd.read_excel('./outputs/primary_agg.xlsx')
osus_ind_name = 'osus'
indicator_type = 'primary'
#INFERENCE
frames2=[]
for company in primary_agg['company'].unique():
   frames = \Pi
   for year in primary_agg['year'].unique():
      company_year = primary_agg[(primary_agg['company'] == company) &
                               (primary_agg['year'] == year)]
       #print(year, company)
       s_indicators = [company_year[company_year[indicator_type] == indicator].
→iloc[0] for indicator in primary_agg[indicator_type].unique()]
       frames.append(InferenceEngine.
→s_p(s_indicators,osus_ind_name,indicator_type)) #apply inference engine
   frames2.append(pd.concat(frames,axis='columns').T)
osus = pd.concat(frames2).reset_index(drop=True)
osus.to_excel('./outputs/osus_fuzz.xlsx')
# -----
print('Defuzzification')
```

```
osus = pd.read_excel('./outputs/osus_fuzz.xlsx')
osus.index= pd.to_datetime(osus['year'].astype(str), format='%Y')
elvllflifhhvheh = Fuzzification.elvllflifhhvheh()
osus['crisp'] = (elvllflifhhvheh['el'].idxmax()*osus['el'] +
                elvllflifhhvheh['vl'].idxmax()*osus['vl'] +
                elvllflifhhvheh['1'].idxmax()*osus['1'] +
                elvllflifhhvheh['fl'].idxmax()*osus['fl'] +
                elvllflifhhvheh['i'].idxmax()*osus['i'] +
                elvllflifhhvheh['fh'].idxmax()*osus['fh'] +
                elvllflifhhvheh['h'].idxmax()*osus['h'] +
                elvllflifhhvheh['vh'].idxmax()*osus['vh'] +
                elvllflifhhvheh['eh'].idxmax()*osus['eh'])/
-osus[['el','vl','l','fl','i','fh','h','vh','eh']].sum(axis='columns')
crisp = osus[['company','crisp']]
crisp.to_excel('./outputs/osus_crisp.xlsx')
fig, ax = plt.subplots(figsize=(8,4))
for idx, gp in crisp.groupby('company'):
   gp.plot(ax=ax,label=idx)
plt.grid(); plt.xlabel('Year'); plt.ylabel('OSUS');plt.legend(crisp.
→groupby('company').indices)
fig.savefig('./outputs/final_output.png', dpi=300, bbox_inches='tight')
print('COMPLETE')
```