## Sensor Schemes CO

November 20, 2021

#### 1 DATA

```
[66]: import pandas as pd
      Ref=pd.read_csv('Ref.csv')
      Ref["CO"] = 1000 * Ref["CO"]
      Ref['Date'] = pd.to_datetime(Ref['Date_Time'])
      Ref=Ref.set_index('Date')
      Ref.drop('Date_Time',axis = 1, inplace = True)
      Ref=Ref.resample('5min').mean()
      Ref=Ref[76463:137376]
      Ref_CO=Ref['CO'].to_list()
      Ref NO2=Ref['NO2'].to list()
      Ref_S02=Ref['S02'].to_list()
      Ref_03=Ref['03'].to_list()
[67]: import random
      import pandas as pd
      import scipy.io
      import numpy as np
      data = pd.read_csv('CO.txt', header = None,low_memory=False)
      data.columns=['WE','AE','Temp','RH','Time']
      Time=data['Time'].to_list()
      time=[]
      for i in range(len(Time)):
          time.append(float(abs(Time[i])))
      Time=np.array(time)
      Date=pd.to_datetime(Time-719529,unit='d').round('s')
      data['Date'] = Date.tolist()
      data=data.set_index('Date')
      data.drop('Time',axis = 1, inplace = True)
      data=data.resample('5min').mean()
      Data_CO=data
      Data_CO['Ref']=Ref_CO
      WE=Data_CO['WE'].to_list()
      AE=Data_CO['AE'].to_list()
      signal=np.array(WE)-np.array(AE)
      Data_CO['Net Signal']=signal
```

```
Data_CO['Month'] = Data_CO.index.month
      Data CO['Day of week'] = Data CO.index.dayofweek
      Data_CO['Day'] = Data_CO.index.day
      Data_CO['Hour'] = Data_CO.index.hour
      CO_Data=Data_CO
      CO_Data=CO_Data[(CO_Data[CO_Data.columns] >= 0).all(axis=1)]
      CO Data=CO Data.dropna()
      data = pd.read_csv('Conc_CO.txt', header = None,low_memory=False)
      data.columns=['Lab1','Temp','RH','Time','Ref']
      Time=data['Time'].to_list()
      time=[]
      for i in range(len(Time)):
          time.append(float(abs(Time[i])))
      Time=np.array(time)
      Date=pd.to_datetime(Time-719529,unit='d').round('s')
      data['Date'] = Date.tolist()
      data=data.set_index('Date')
      data.drop('Time',axis = 1, inplace = True)
      data=data.resample('5min').mean()
      Data_CO=data
      signal=np.array(WE)-np.array(AE)
      Data CO['Net Signal']=signal
      Data_CO['Month'] = Data_CO.index.month
      Data CO['Day of week'] = Data CO.index.dayofweek
      Data_CO['Day'] = Data_CO.index.day
      Data_CO['Hour'] = Data_CO.index.hour
      CO Data=Data CO
      CO_Data=CO_Data.resample('5min').mean()
      CO_Data=CO_Data[(CO_Data[CO_Data.columns] >= 0).all(axis=1)]
      CO_Data=CO_Data.dropna()
      CO_Data.shape
[67]: (45117, 9)
[68]: CO_Data=CO_Data.resample('h').mean()
      CO_Data=CO_Data.dropna()
[69]: def MBE(true, pred):
          true=np.array(true)
          pred=np.array(pred)
          mbe=np.mean(true-pred)
          return mbe
      def CRMSE(true,pred):
          true=np.array(true)
          pred=np.array(pred)
          crmse=np.sqrt(np.mean(((true-np.mean(true))-(pred-np.mean(pred)))**2))
          if np.std(pred)>np.std(true):
```

```
crmse=crmse
else:
    crmse=-crmse
return crmse
```

```
[70]: df1=[x for _, x in CO_Data.groupby('Month')]
      data_oct=df1[4]
      \#data\_oct=data\_oct.sample(frac=1)
      data_nov=df1[5]
      #data_nov=data_nov.sample(frac=1)
      data dec=df1[6]
      #data_dec=data_dec.sample(frac=1)
      data_jan=df1[0]
      \#data_jan=data_jan.sample(frac=1)
      data_feb=df1[1]
      \#data\_feb=data\_feb.sample(frac=1)
      data_mar=df1[2]
      #data_mar=data_mar.sample(frac=1)
      data_apr=df1[3]
      \#data\_apr=data\_apr.sample(frac=1)
      data=[data_oct,data_nov,data_dec,data_jan,data_feb,data_mar]
      data_apr.head()
```

[70]:	Lab1	Temp	RH	Ref	Net Signal \
Date					
2020-04-01 00:00:00	229.616988	13.851747	82.302940	197.721703	95.726012
2020-04-01 01:00:00	199.698257	13.454867	83.531290	263.747572	82.533367
2020-04-01 02:00:00	171.861486	13.304556	82.558732	253.751264	69.676736
2020-04-01 03:00:00	136.027044	13.187792	82.926804	204.773281	52.308946
2020-04-01 04:00:00	150.059790	13.625268	84.595291	217.311244	58.030088
	Month Day_	of_week Da	y Hour		
Date					
2020-04-01 00:00:00	4.0	2.0 1.	0.0		
2020-04-01 01:00:00	4.0	2.0 1.	0 1.0		
2020-04-01 02:00:00	4.0	2.0 1.	0 2.0		
2020-04-01 03:00:00	4.0	2.0 1.	0 3.0		
2020-04-01 04:00:00	4.0	2.0 1.	0 4.0		

# 2 Mothly schemes

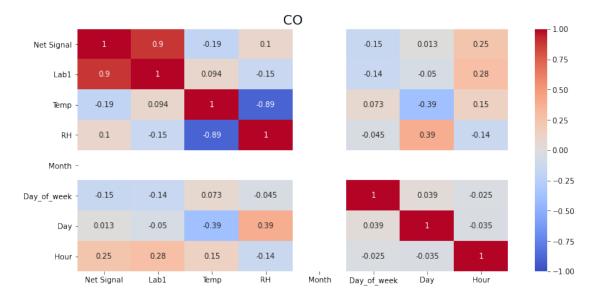
## 3 Oct 2019

```
[71]: from sklearn.model_selection import train_test_split from sklearn.linear_model import LinearRegression import sklearn.metrics as sm
```

```
import matplotlib.pyplot as plt
```

```
[72]: #,'Month','Day_of_week','Day','Hour'
X=data_oct[['Net Signal','Lab1','Temp','RH','Month','Day_of_week','Day','Hour']]
y=data_oct['Ref']
A=np.array(y)/np.mean(y)
A=sorted(A, reverse=True)
sum1=sum(A[:100])
mean1=np.std(y)/np.mean(y)
```

#### [73]: Text(0.5, 1.0, 'CO')



#### 4 R.F.

```
[74]: from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression

#from sklearn.metrics import mean_absolute_error as mae

#from sklearn.metrics import mean_absolute_percentage_error

import sklearn.metrics as sm

import matplotlib.pyplot as plt

import numpy as np
```

```
from sklearn.ensemble import RandomForestRegressor
       # create regressor object
      regressor = RandomForestRegressor(n_estimators = 500,min_samples_split=_
       →2,min_samples_leaf= 1,max_features= 'sqrt',
                                        random state = |
       →0, max depth=None, bootstrap=False)
[75]: Day=[2*i for i in range(1,8)]
      Rmse1_rf=[]
      RMSE1 rf=[]
      for i in range(1,8):
          regressor.fit(X[:48*i].drop(['Lab1'], axis=1), y[:48*i])
          pred=regressor.predict(X.drop(['Lab1'], axis=1))
          mse=round(sm.r2_score(y_test, pred), 2)
          #mse=round(np.sqrt(sm.mean squared error(y test, pred))/np.mean(y test),2)
          #rmse= mape=round(mean_absolute_percentage_error(y_test,pred),2)
          rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
          Rmse1 rf.append(mse)
          RMSE1_rf.append(rmse)
[76]: A=y.to_list()
      Ext oct=[]
      for i in range(len(A)):
          if A[i]>3*np.mean(A):
              Ext_oct.append(i)
      N_Ext_oct=len(Ext_oct)
      N_Ext_oct
[76]: 13
```

```
[77]: mean_oct=np.mean(y)
N_oct=y.shape[0]
Mean_Rmse_oct=np.mean(Rmse1_rf)
```

from xgboost import XGBRegressor from numpy import absolute from pandas import read\_csv from sklearn.model\_selection import cross\_val\_score from sklearn.model\_selection import RepeatedKFold # create an xgboost regression model #n\_estimators=10000, max\_depth=5, eta=0.01, subsample=0.9,colsample\_bytree=0.4,alpha=10 model = XG-BRegressor(n\_estimators=10000, max\_depth=5, eta=0.01, subsample=0.9, colsample\_bytree=0.4,alpha=10)

 $Rmse1\_ann=[] RMSE1\_ann=[] for i in range(1,8): model.fit(X[:48*i].drop([`Lab1'], axis=1), y[:48*i]) pred=model.predict(X.drop([`Lab1'], axis=1)) y\_test=y mse=round(np.sqrt(sm.mean\_squared\_error(y\_test, pred))/np.mean(y\_test),2) #rmse=$ 

mape=round(mean\_absolute\_percentage\_error(y\_test,pred),2) rmse=round(np.corrcoef(y\_test,pred)[0, 1],2) Rmse1\_ann.append(mse) RMSE1\_ann.append(rmse)

#### 6 Nov 2019

# RF

```
[78]: X=data_nov[['Net Signal','Lab1','Temp','RH','Month','Day_of_week','Day','Hour']]
      y=data nov['Ref']
      A=np.array(y)/np.mean(y)
      A=sorted(A, reverse=True)
      sum2=sum(A[:100])
      mean2=np.std(y)/np.mean(y)
      Rmse2 rf=[]
      RMSE2_rf=[]
      for i in range (1,8):
          regressor.fit(X[:48*i].drop(['Lab1'], axis=1), y[:48*i])
          pred=regressor.predict(X.drop(['Lab1'], axis=1))
          y_test=y
          mse=round(sm.r2_score(y_test, pred), 2)
          #mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
          #rmse= mape=round(mean absolute percentage error(y test,pred),2)
          rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
          Rmse2_rf.append(mse)
          RMSE2_rf.append(rmse)
[79]: A=y.to_list()
      Ext_nov=[]
      for i in range(len(A)):
          if A[i]>3*np.mean(A):
              Ext_nov.append(i)
      N_Ext_nov=len(Ext_nov)
      N_Ext_nov
[79]: 26
[80]: mean_nov=np.mean(y)
      N_nov=y.shape[0]
      Mean_Rmse_nov=np.mean(Rmse2_rf)
```

#### 7 XGBoost

```
Rmse2_ann=[] RMSE2_ann=[] for i in range(1,8): model.fit(X[:48*i].drop(['Lab1'], axis=1), y[:48*i]) pred=model.predict(X.drop(['Lab1'], axis=1)) y_test=y

mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)

#rmse= mape=round(mean_absolute_percentage_error(y_test,pred),2)

rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
```

```
Rmse2_ann.append(mse)
RMSE2_ann.append(rmse)
```

#### 8 Dec 2019

#### 9 RF

```
[81]: X=data_dec[['Net Signal','Lab1','Temp','RH','Month','Day_of_week','Day','Hour']]
      y=data_dec['Ref']
      A=np.array(y)/np.mean(y)
      A=sorted(A, reverse=True)
      sum3=sum(A[:100])
      mean3=np.std(y)/np.mean(y)
      Rmse3_rf=[]
      RMSE3_rf=[]
      for i in range(1,8):
          regressor.fit(X[:48*i].drop(['Lab1'], axis=1), y[:48*i])
          pred=regressor.predict(X.drop(['Lab1'], axis=1))
          mse=round(sm.r2_score(y_test, pred), 2)
          #mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
          #rmse= mape=round(mean_absolute_percentage_error(y_test,pred),2)
          rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
          Rmse3_rf.append(mse)
          RMSE3_rf.append(rmse)
[82]: A=y.to_list()
      Ext_dec=[]
      for i in range(len(A)):
          if A[i]>3*np.mean(A):
              Ext_dec.append(i)
      N_Ext_dec=len(Ext_dec)
      N_Ext_dec
[82]: 20
[83]: mean_dec=np.mean(y)
      N_dec=y.shape[0]
      Mean_Rmse_dec=np.mean(Rmse3_rf)
```

#### 10 XGBoost

 $Rmse3\_ann=[] RMSE3\_ann=[] for i in range(1,8): model.fit(X[:48*i].drop([`Lab1'], axis=1), y[:48*i]) pred=model.predict(X.drop([`Lab1'], axis=1)) y\_test=y mse=round(np.sqrt(sm.mean\_squared\_error(y\_test, pred))/np.mean(y\_test),2) #rmse=mape=round(mean\_absolute\_percentage\_error(y\_test,pred),2) rmse=round(np.corrcoef(y\_test,pred))[0, 1],2) Rmse3 ann.append(mse) RMSE3 ann.append(rmse)$ 

#### 11 Jan 2020

#### 12 RF

```
[84]: X=data_jan[['Net Signal', 'Lab1', 'Temp', 'RH', 'Month', 'Day_of_week', 'Day', 'Hour']]
      y=data_jan['Ref']
      A=np.array(y)/np.mean(y)
      A=sorted(A, reverse=True)
      sum4=sum(A[:100])
      mean4=np.std(y)/np.mean(y)
      Rmse4_rf=[]
      RMSE4 rf=[]
      for i in range(1,8):
          regressor.fit(X[:48*i].drop(['Lab1'], axis=1), y[:48*i])
          pred=regressor.predict(X.drop(['Lab1'], axis=1))
          v test=v
          mse=round(sm.r2_score(y_test, pred), 2)
          \#mse=round(np.sqrt(sm.mean\_squared\_error(y\_test, pred))/np.mean(y\_test),2)
          #rmse= mape=round(mean_absolute_percentage_error(y_test,pred),2)
          rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
          Rmse4_rf.append(mse)
          RMSE4_rf.append(rmse)
[85]: A=y.to_list()
      Ext jan=[]
      for i in range(len(A)):
          if A[i]>3*np.mean(A):
              Ext_jan.append(i)
      N_Ext_jan=len(Ext_jan)
      N_Ext_jan
[85]: 23
[86]: mean_jan=np.mean(y)
      N_jan=y.shape[0]
      Mean_Rmse_jan=np.mean(Rmse4_rf)
```

#### 13 XGBoost

```
[87]: Rmse4_ann=[]
RMSE4_ann=[]
for i in range(1,8):
    model.fit(X[:48*i].drop(['Lab1'], axis=1), y[:48*i])
    pred=model.predict(X.drop(['Lab1'], axis=1))
    y_test=y
    mse=round(sm.r2_score(y_test, pred), 2)
    #mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
```

```
#rmse= mape=round(mean_absolute_percentage_error(y_test,pred),2)
rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
Rmse4_ann.append(mse)
RMSE4_ann.append(rmse)
```

#### 14 Feb 2020

#### 15 RF

```
X=data feb[['Net
                              Signal', 'Lab1', 'Temp', 'RH', 'Month', 'Day of week', 'Day', 'Hour']
v=data feb['Ref']
                        A=np.array(y)/np.mean(y)
                                                         A = sorted(A,
                                                                             reverse=True)
sum5=sum(A[:100])
                        mean5 = np.std(y)/np.mean(y)
                                                           Rmse5 rf=[]
                                                                              RMSE5 rf=[]
                                 regressor.fit(X[:576*i].drop(['Lab1'],
               range(1,11):
                                                                      axis=1),
                                                                                  y[:576*i]
pred=regressor.predict(X[576*i:].drop(['Lab1'],
                                                        axis=1)
                                                                           y_test=y[576*i:]
mse=round(np.sqrt(sm.mean_squared_error(y_test,
                                                     pred))/np.mean(y_test),2)
                                                                                   #rmse=
mape=round(mean_absolute_percentage_error(y_test,pred),2) rmse=round(np.corrcoef(y_test,
pred)[0, 1],2) Rmse5_rf.append(mse) RMSE5_rf.append(rmse)
A=y.to_list() Ext_feb=[] for i in range(len(A)): if A[i]>3*np.mean(A): Ext_feb.append(i)
N_Ext_feb=len(Ext_feb) N_Ext_feb
mean_feb=np.mean(y) N_feb=y.shape[0] Mean_Rmse_feb=np.mean(Rmse5_rf)
```

#### 16 XGBoost

#### 17 March 2020

#### 18 RF

```
[88]: X=data_mar[['Net Signal','Lab1','Temp','RH','Month','Day_of_week','Day','Hour']]
y=data_mar['Ref']
A=np.array(y)/np.mean(y)
A=sorted(A, reverse=True)
sum6=sum(A[:100])
mean6=np.std(y)/np.mean(y)
Rmse6_rf=[]
RMSE6_rf=[]
for i in range(1,8):
```

```
regressor.fit(X[:48*i].drop(['Lab1'], axis=1), y[:48*i])
pred=regressor.predict(X.drop(['Lab1'], axis=1))
y_test=y
mse=round(sm.r2_score(y_test, pred), 2)
#mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
#rmse= mape=round(mean_absolute_percentage_error(y_test,pred),2)
rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
Rmse6_rf.append(mse)
RMSE6_rf.append(rmse)
```

```
[89]: A=y.to_list()
Ext_mar=[]
for i in range(len(A)):
    if A[i]>3*np.mean(A):
        Ext_mar.append(i)
N_Ext_mar=len(Ext_mar)
N_Ext_mar
```

#### [89]: 11

```
[90]: mean_mar=np.mean(y)
    N_mar=y.shape[0]
    Mean_Rmse_mar=np.mean(Rmse6_rf)
```

#### 19 XGBoost

```
Rmse6_ann=[] RMSE6_ann=[] for i in range(1,8): model.fit(X[:48*i].drop(['Lab1'], axis=1),
y[:48*i]) pred=model.predict(X.drop(['Lab1'], axis=1)) y_test=y

mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
#rmse= mape=round(mean_absolute_percentage_error(y_test,pred),2)
rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
Rmse6_ann.append(mse)
RMSE6_ann.append(rmse)
```

## 20 April 2020

#### 21 R.F

```
Signal', 'Lab1', 'Temp', 'RH', 'Month', 'Day of week', 'Day', 'Hour']]
X=data apr[['Net
y=data apr['Ref']
                         A=np.array(y)/np.mean(y)
                                                           A = sorted(A,
                                                                               reverse=True)
                                                           Rmse10 rf=[]
sum7=sum(A[:100])
                        mean7=np.std(y)/np.mean(y)
                                                                              RMSE10 rf=[]
               range(1,11):
                                 regressor.fit(X[:576*i].drop(['Lab1'],
                                                                                    y[:576*i])
                                                                       axis=1),
pred=regressor.predict(X[576*i:].drop(['Lab1'],
                                                         axis=1)
                                                                            y test=y[576*i:]
mse=round(np.sqrt(sm.mean_squared_error(y_test,
                                                      pred))/np.mean(y_test),2)
                                                                                    #rmse=
mape=round(mean_absolute_percentage_error(y_test,pred),2) rmse=round(np.corrcoef(y_test,
pred)[0, 1],2) Rmse10_rf.append(mse) RMSE10_rf.append(rmse)
```

```
A=y.to_list() Ext_apr=[] for i in range(len(A)): if A[i]>3*np.mean(A): Ext_apr.append(i) N_Ext_apr=len(Ext_apr) N_Ext_apr
```

 $\label{eq:Rmse10_ann} RMSE10\_ann=[] for i in range(1,11): model.fit(X[:576*i].drop(['Lab1'], axis=1), y[:576*i]) pred=model.predict(X[576*i:].drop(['Lab1'], axis=1)) y_test=y[576*i:] mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2) \#rmse=mape=round(mean_absolute_percentage_error(y_test,pred),2) rmse=round(np.corrcoef(y_test,pred)[0, 1],2) Rmse10_ann.append(mse) RMSE10_ann.append(rmse) mean_apr=np.mean(y) N_apr=y.shape[0] Mean_Rmse_apr=np.mean(Rmse10_rf) \\$ 

Mean\_Rmse=[Mean\_Rmse\_oct,Mean\_Rmse\_nov,Mean\_Rmse\_dec,Mean\_Rmse\_jan,Mean\_Rmse\_feb,Mean\_Mean\_conc=[mean\_oct,mean\_nov,mean\_dec,mean\_jan,mean\_feb,mean\_mar,mean\_apr]
N=[N\_oct,N\_nov,N\_dec,N\_jan,N\_feb,N\_mar,N\_apr] N\_Ext=[N\_Ext\_oct,N\_Ext\_nov,N\_Ext\_dec,N\_Ext\_dec,N\_Ext\_dec,N\_ext\_dec,N\_

 $Rmse=[] \ for \ i \ in \ range(10): \ A=[Rmse1\_rf[i],Rmse2\_rf[i],Rmse3\_rf[i],Rmse4\_rf[i],Rmse5\_rf[i],Rmse6\_rf[i],Rmse6\_rf[i],Rmse10\_rmse10] \\ Rmse.append(A) \ RMSE=Rmse[0]+Rmse[1]+Rmse[2]+Rmse[3]+Rmse[4]+Rmse[5]+Rmse[6]+Rmse[6]+Rmse[7] \\ Conc=Mean\_conc+M$ 

```
[91]: import plotly.express as px
      from IPython.display import Image
      import plotly.graph_objects as go
      import numpy as np
      Day_1_rf=[RMSE1_rf[0],RMSE2_rf[0],RMSE3_rf[0],RMSE4_rf[0],RMSE6_rf[0]]
      Day_2_rf=[RMSE1_rf[1],RMSE2_rf[1],RMSE3_rf[1],RMSE4_rf[1],RMSE6_rf[1]]
      Day_3_rf=[RMSE1_rf[2],RMSE2_rf[2],RMSE3_rf[2],RMSE4_rf[2],RMSE6_rf[2]]
      Day_4_rf=[RMSE1_rf[3],RMSE2_rf[3],RMSE3_rf[3],RMSE4_rf[3],RMSE6_rf[3]]
      Day_5_rf=[RMSE1_rf[4],RMSE2_rf[4],RMSE3_rf[4],RMSE4_rf[4],RMSE6_rf[4]]
      Day_6_rf=[RMSE1_rf[5],RMSE2_rf[5],RMSE3_rf[5],RMSE4_rf[5],RMSE6_rf[5]]
      Day_7_rf=[RMSE1_rf[6],RMSE2_rf[6],RMSE3_rf[6],RMSE4_rf[6],RMSE6_rf[6]]
      \#Day\_8\_rf = [RMSE1\_rf[7], RMSE2\_rf[7], RMSE3\_rf[7], RMSE4\_rf[7], RMSE6\_rf[7]]
      \#Day_9 rf = [RMSE1_rf[8], RMSE2_rf[8], RMSE3_rf[8], RMSE4_rf[8], RMSE6_rf[8]]
      \#Day_10_rf = [RMSE1_rf[9], RMSE2_rf[9], RMSE3_rf[9], RMSE4_rf[9], RMSE4_rf[9]]
      Day_1_ann=[RMSE1_ann[0],RMSE2_ann[0],RMSE3_ann[0],RMSE4_ann[0],RMSE6_ann[0]]
      Day_2_ann=[RMSE1_ann[1],RMSE2_ann[1],RMSE3_ann[1],RMSE4_ann[1],RMSE6_ann[1]]
      Day_3_ann=[RMSE1_ann[2],RMSE2_ann[2],RMSE3_ann[2],RMSE4_ann[2],RMSE6_ann[2]]
      Day_4_ann=[RMSE1_ann[3],RMSE2_ann[3],RMSE3_ann[3],RMSE4_ann[3],RMSE6_ann[3]]
      Day_5_ann=[RMSE1_ann[4],RMSE2_ann[4],RMSE3_ann[4],RMSE4_ann[4],RMSE6_ann[4]]
      Day_6_ann=[RMSE1_ann[5],RMSE2_ann[5],RMSE3_ann[5],RMSE4_ann[5],RMSE6_ann[5]]
      Day_7_ann=[RMSE1_ann[6],RMSE2_ann[6],RMSE3_ann[6],RMSE4_ann[6],RMSE6_ann[6]]
      #Day 8 ann=[RMSE1 ann[7], RMSE2 ann[7], RMSE3 ann[7], RMSE4 ann[7], RMSE6 ann[7]]
      #Day_9_ann=[RMSE1_ann[8],RMSE2_ann[8],RMSE3_ann[8],RMSE4_ann[8],RMSE6_ann[8]]
      #Day 10 ann=[RMSE1 ann[9],RMSE2 ann[9],RMSE3 ann[9],RMSE4 ann[9],RMSE6 ann[9]]
```

```
Day_2_RF=[Rmse1_rf[1],Rmse2_rf[1],Rmse3_rf[1],Rmse4_rf[1],Rmse6_rf[1]]
      Day 3 RF=[Rmse1 rf[2],Rmse2 rf[2],Rmse3 rf[2],Rmse4 rf[2],Rmse6 rf[2]]
      Day_4_RF=[Rmse1_rf[3],Rmse2_rf[3],Rmse3_rf[3],Rmse4_rf[3],Rmse6_rf[3]]
      Day 5 RF=[Rmse1 rf[4],Rmse2 rf[4],Rmse3 rf[4],Rmse4 rf[4],Rmse6 rf[4]]
      Day_6_RF=[Rmse1_rf[5],Rmse2_rf[5],Rmse3_rf[5],Rmse4_rf[5],Rmse6_rf[5]]
      Day_7_RF=[Rmse1_rf[6],Rmse2_rf[6],Rmse3_rf[6],Rmse4_rf[6],Rmse6_rf[6]]
      \#Day_{RF}=[Rmse1\_rf[7],Rmse2\_rf[7],Rmse3\_rf[7],Rmse4\_rf[7],Rmse6\_rf[7]]
      \#Day_9 = RF = [Rmse_1 rf[8], Rmse_2 rf[8], Rmse_3 rf[8], Rmse_4 rf[8], Rmse_6 rf[8]]
      \#Day_10_RF = [Rmse1_rf[9], Rmse2_rf[9], Rmse3_rf[9], Rmse4_rf[9], Rmse6_rf[9]]
      Day 1 ANN=[Rmse1 ann[0], Rmse2 ann[0], Rmse3 ann[0], Rmse4 ann[0], Rmse6 ann[0]]
      Day 2 ANN=[Rmse1 ann[1], Rmse2 ann[1], Rmse3 ann[1], Rmse4 ann[1], Rmse6 ann[1]]
      Day 3 ANN=[Rmse1 ann[2], Rmse2 ann[2], Rmse3 ann[2], Rmse4 ann[2], Rmse6 ann[2]]
      Day_4_ANN=[Rmse1_ann[3],Rmse2_ann[3],Rmse3_ann[3],Rmse4_ann[3],Rmse6_ann[3]]
      Day 5 ANN=[Rmse1 ann[4], Rmse2 ann[4], Rmse3 ann[4], Rmse4 ann[4], Rmse6 ann[4]]
      Day 6 ANN=[Rmse1 ann [5], Rmse2 ann [5], Rmse3 ann [5], Rmse4 ann [5], Rmse6 ann [5]]
      Day 7 ANN=[Rmse1 ann[6], Rmse2 ann[6], Rmse3 ann[6], Rmse4 ann[6], Rmse6 ann[6]]
      \#Day_{ann}[7], Rmse1_{ann}[7], Rmse2_{ann}[7], Rmse4_{ann}[7], Rmse4_{ann}[7], Rmse6_{ann}[7]
      \#Day\_9\_ANN=[Rmse1\_ann[8],Rmse2\_ann[8],Rmse3\_ann[8],Rmse4\_ann[8],Rmse6\_ann[8]]
      \#Day\_10\_ANN = [Rmse1\_ann[9], Rmse2\_ann[9], Rmse3\_ann[9], Rmse4\_ann[9], Rmse6\_ann[9]]
[93]: RF_P=Day_1_rf+Day_2_rf+Day_3_rf+Day_4_rf+Day_5_rf+Day_6_rf+Day_7_rf#+Day_8_rf+Day_9_rf+Day_10_
      #ANN_P=Day_1_ann+Day_2_ann+Day_3_ann+Day_4_ann+Day_5_ann+Day_6_ann+Day_7_ann#+Day_8_ann+Day_9
      RF_R=Day_1_RF+Day_2_RF+Day_3_RF+Day_4_RF+Day_5_RF+Day_6_RF+Day_7_RF#+Day_8_RF+Day_9_RF+Day_10_
      #ANN_R=Day_1_ANN+Day_2_ANN+Day_3_ANN+Day_4_ANN+Day_5_ANN+Day_6_ANN+Day_7_ANN#+Day_8_ANN+Day_9_
      x1=['2' for i in range(5)]
      x2=['4' \text{ for i in } range(5)]
      x3=['6' for i in range(5)]
      x4=['8' for i in range(5)]
      x5=['10' for i in range(5)]
      x6=['12' for i in range(5)]
      x7=['14' for i in range(5)]
      #x8=['16' for i in range(5)]
      #x9=['18' for i in range(5)]
      #x10=['20' for i in range(5)]
      x=x1+x2+x3+x4+x5+x6+x7#+x8+x9+x10
      len(x)
```

[92]: Day 1 RF=[Rmse1 rf[0],Rmse2 rf[0],Rmse3 rf[0],Rmse4 rf[0],Rmse6 rf[0]]

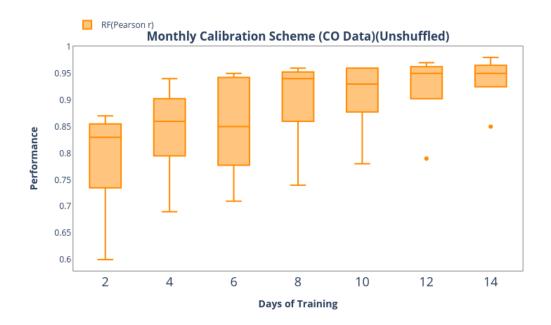
[93]: 35

```
[96]: import chart_studio
      fig = go.Figure()
      a = np.array([1, 2, 3, 4, 5])
      b = np.array([1, 3, 2, 3, 1])
      # Defining x axis
      x = x
      fig.add_trace(go.Box(
          # defining y axis in corresponding
          # to x-axis
          y=RF P,
          x=x,
          name='RF(Pearson r)',
          marker_color='darkorange',
          showlegend=True
      ))
      #fig.add_trace(go.Box(
          #y=ANN_P,
         \# x=x,
          #name='XGBoost(Pearson r)',
          #marker_color='darkcyan',
          #showlegend=True
      #))
      #fig.add_trace(go.Box(
          # defining y axis in corresponding
         # to x-axis
         # y=RF_R,
          \#x=x,
          #name = 'RF(R^2)',
          #marker_color='teal',
          #showlegend=True
      #))
      #fig.add_trace(go.Box(
         # y=ANN_R,
          \#x=x,
          #name='XGBoost(NSRMSE)',
          #marker_color='deeppink',
          #showlegend=True
      #))
```

```
fig.update_layout(autosize=False,
                 title={'text': "<b>Monthly Calibration Scheme (CO__
→Data) (Unshuffled)</b>",
        'y':0.84,
        'x':0.5,
        'xanchor': 'center',
        'yanchor': 'top'},
    width=800,
    height=500,
  legend=dict( yanchor="bottom",
    y=1.05,
    orientation="h"
),
    # group together boxes of the different
    # traces for each value of x
    boxmode='group',
                  plot_bgcolor='rgba(0,0,0,0)'
)
fig.update_xaxes(title_text="<b>Days of Training</b>",tickfont =_

dict(size=18),linewidth=0.5, linecolor='black',
                 mirror=True)
fig.update_vaxes(title_text="<b>Performance</b>",linewidth=0.5,__
⇔linecolor='black',
                 mirror=True)
fig.show()
chart_studio.plotly.sign_in('vinylango', 'gybbJVWfRSUoTcRRSa6J')
chart_studio.plotly.image.save_as(fig, filename='models_boxplot.png')
Image('models_boxplot.png')
```

Г96**]** :



Metric1=['RF' for i in range(len(RF))] Metric2=['XGBoost' for i in range(len(ANN))] Model=Metric1+Metric2 Training=x+x Values=RF+ANN len(Values)

 $\label{eq:width} $\# \text{Violin plot which also show the density of the distribution import plotly.express as px Metric1=['RF' for i in range(len(RF))] Metric2=['XGBoost' for i in range(len(ANN))] Model=Metric1+Metric2 Training=x+x Values=RF+ANN lst=[[Training[i],Values[i],Model[i]] for i in range(len(Model))] df = pd.DataFrame(lst, columns =['Training Days', 'Pearson correlation (r)','Model'])$ 

#fig = px.violin( df,y="Performance", x="Calibration Model", color='Metric', box=True,points="all", #hover\_data=df.columns) fig = px.violin( df,y="Pearson correlation (r)", x="Training Days", color='Model', box=True, hover\_data=df.columns)

fig.update\_layout(autosize=False, width=900, height=500) fig.show() #chart\_studio.plotly.sign\_in('vinylango', 'gybbJVWfRSUoTcRRSa6J') #chart\_studio.plotly.image.save\_as(fig, filename='models\_violinplots.png') #Image('models\_violinplots.png')

#### 23 Seasonal Calibration Scheme

#### 24 Fall

#### 25 RF

```
[52]: frame1=[data_oct,data_nov]
      fall=pd.concat(frame1)
      #fall=fall.sample(frac=1)
      Day=[5*i for i in range(1,11) ]
      X=fall[['Net Signal','Lab1','Temp','RH','Month','Day of week','Day','Hour']]
      y=fall['Ref']
      mean1=np.std(y)
      Rmse7_rf=[]
      RMSE7_rf=[]
      for i in range(1,11):
          regressor.fit(X[:120*i].drop(['Lab1'], axis=1), y[:120*i])
          pred=regressor.predict(X.drop(['Lab1'], axis=1))
          y test=y
          mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
          rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
          Rmse7 rf.append(mse)
          RMSE7_rf.append(rmse)
```

#### 26 XGBoost

```
[53]: Rmse7_ann=[]
RMSE7_ann=[]
for i in range(1,11):
    model.fit(X[:120*i].drop(['Lab1'], axis=1), y[:120*i])
    pred=model.predict(X.drop(['Lab1'], axis=1))
    y_test=y
    mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
    #rmse= mape=round(mean_absolute_percentage_error(y_test,pred),2)
    rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
    Rmse7_ann.append(mse)
    RMSE7_ann.append(rmse)
```

#### 27 Winter

#### 28 R.F

```
[54]: frame1=[data_dec,data_jan,data_feb]
winter=pd.concat(frame1)
#winter=winter.sample(frac=1)
X=winter[['Net Signal','Lab1','Temp','RH','Month','Day_of_week','Day','Hour']]
```

```
y=winter['Ref']
mean2=np.std(y)
Rmse8_rf=[]
RMSE8_rf=[]
for i in range(1,11):
    regressor.fit(X[:120*i].drop(['Lab1'], axis=1), y[:120*i])
    pred=regressor.predict(X.drop(['Lab1'], axis=1))
    y_test=y
    mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
    rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
    Rmse8_rf.append(mse)
    RMSE8_rf.append(rmse)
```

```
Rmse8_ann=[]
RMSE8_ann=[]
for i in range(1,11):
    model.fit(X[:120*i].drop(['Lab1'], axis=1), y[:120*i])
    pred=model.predict(X.drop(['Lab1'], axis=1))
    y_test=y
    mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
    #rmse= mape=round(mean_absolute_percentage_error(y_test,pred),2)
    rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
    Rmse8_ann.append(mse)
    RMSE8_ann.append(rmse)
```

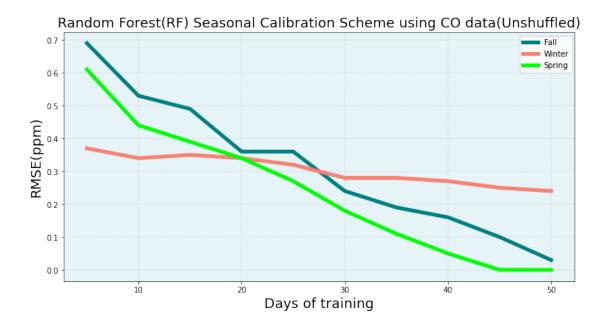
## 30 Spring

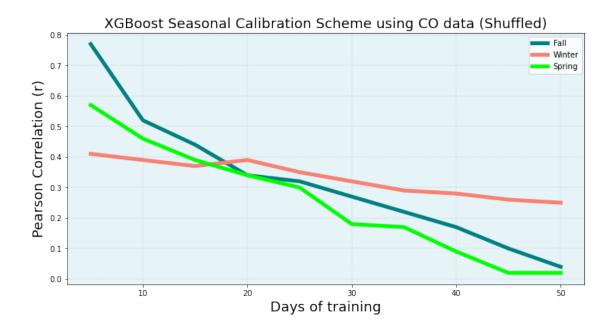
#### 31 RF

```
[56]: frame1=[data_mar,data_apr]
    spring=pd.concat(frame1)
    #spring=spring.sample(frac=1)

X=spring[['Net Signal','Lab1','Temp','RH','Month','Day_of_week','Day','Hour']]
    y=spring['Ref']
    mean3=np.std(y)
    Rmse9_rf=[]
    RMSE9_rf=[]
    for i in range(1,11):
        regressor.fit(X[:120*i].drop(['Lab1'], axis=1), y[:120*i])
        pred=regressor.predict(X.drop(['Lab1'], axis=1))
        y_test=y
        mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
        rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
        Rmse9_rf.append(mse)
```

```
[57]: Rmse9_ann=[]
      RMSE9_ann=[]
      for i in range(1,11):
          model.fit(X[:120*i].drop(['Lab1'], axis=1), y[:120*i])
          pred=model.predict(X.drop(['Lab1'], axis=1))
          y_test=y
          mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
          #rmse= mape=round(mean_absolute_percentage_error(y_test,pred),2)
          rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
          Rmse9 ann.append(mse)
          RMSE9_ann.append(rmse)
[58]: fig= plt.figure(figsize=(12,6))
      ax = fig.add_subplot(111)
      ax.patch.set_facecolor('lightblue')
      ax.patch.set_alpha(0.3)
      plt.plot(Day,Rmse7_rf,color='teal',linewidth=5)
      plt.plot(Day,Rmse8_rf,color='salmon',linewidth=5)
      plt.plot(Day,Rmse9_rf,color='lime',linewidth=5)
```



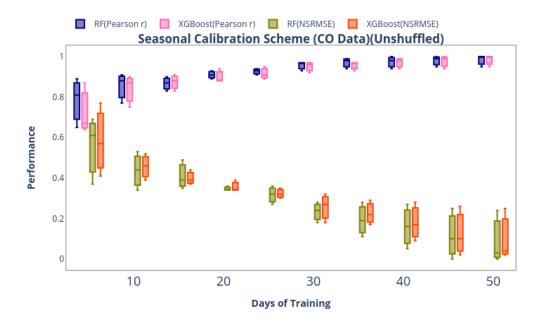


```
[60]: #import chart studio
      import plotly.express as px
      from IPython.display import Image
      import plotly.graph_objects as go
      import numpy as np
      Day 1 rf=[RMSE7 rf[0],RMSE8 rf[0],RMSE9 rf[0]]
      Day_2_rf=[RMSE7_rf[1],RMSE8_rf[1],RMSE9_rf[1]]
      Day_3_rf=[RMSE7_rf[2],RMSE8_rf[2],RMSE9_rf[2]]
      Day_4_rf=[RMSE7_rf[3],RMSE8_rf[3],RMSE9_rf[3]]
      Day_5_rf=[RMSE7_rf[4],RMSE8_rf[4],RMSE9_rf[4]]
      Day_6_rf=[RMSE7_rf[5],RMSE8_rf[5],RMSE9_rf[5]]
      Day_7_rf=[RMSE7_rf[6],RMSE8_rf[6],RMSE9_rf[6]]
      Day_8_rf=[RMSE7_rf[7],RMSE8_rf[7],RMSE9_rf[7]]
      Day 9 rf=[RMSE7 rf[8],RMSE8 rf[8],RMSE9 rf[8]]
      Day_10_rf=[RMSE7_rf[9],RMSE8_rf[9],RMSE9_rf[9]]
      Day 1 ann=[RMSE7 ann[0],RMSE8 ann[0],RMSE9 ann[0]]
      Day_2_ann=[RMSE7_ann[1],RMSE8_ann[1],RMSE9_ann[1]]
      Day_3_ann=[RMSE7_ann[2],RMSE8_ann[2],RMSE9_ann[2]]
      Day_4_ann=[RMSE7_ann[3],RMSE8_ann[3],RMSE9_ann[3]]
      Day_5_ann=[RMSE7_ann[4],RMSE8_ann[4],RMSE9_ann[4]]
      Day_6_ann=[RMSE7_ann[5],RMSE8_ann[5],RMSE9_ann[5]]
      Day_7_ann=[RMSE7_ann[6],RMSE8_ann[6],RMSE9_ann[6]]
      Day_8_ann=[RMSE7_ann[7],RMSE8_ann[7],RMSE9_ann[7]]
      Day_9_ann=[RMSE7_ann[8],RMSE8_ann[8],RMSE9_ann[8]]
      Day_10_ann=[RMSE7_ann[9],RMSE8_ann[9],RMSE9_ann[9]]
```

```
[61]: Day_1_RF=[Rmse7_rf[0],Rmse8_rf[0],Rmse9_rf[0]]
            Day_2_RF=[Rmse7_rf[1],Rmse8_rf[1],Rmse9_rf[1]]
            Day_3_RF=[Rmse7_rf[2],Rmse8_rf[2],Rmse9_rf[2]]
            Day_4_RF=[Rmse7_rf[3],Rmse8_rf[3],Rmse9_rf[3]]
            Day_5_RF=[Rmse7_rf[4],Rmse8_rf[4],Rmse9_rf[4]]
            Day_6_RF=[Rmse7_rf[5],Rmse8_rf[5],Rmse9_rf[5]]
            Day_7_RF=[Rmse7_rf[6],Rmse8_rf[6],Rmse9_rf[6]]
            Day_8_RF=[Rmse7_rf[7],Rmse8_rf[7],Rmse9_rf[7]]
            Day_9_RF=[Rmse7_rf[8],Rmse8_rf[8],Rmse9_rf[8]]
            Day_10_RF=[Rmse7_rf[9],Rmse8_rf[9],Rmse9_rf[9]]
            Day_1_ANN=[Rmse7_ann[0],Rmse8_ann[0],Rmse9_ann[0]]
            Day_2_ANN=[Rmse7_ann[1],Rmse8_ann[1],Rmse9_ann[1]]
            Day_3_ANN=[Rmse7_ann[2],Rmse8_ann[2],Rmse9_ann[2]]
            Day_4_ANN=[Rmse7_ann[3],Rmse8_ann[3],Rmse9_ann[3]]
            Day_5_ANN=[Rmse7_ann[4],Rmse8_ann[4],Rmse9_ann[4]]
            Day_6_ANN=[Rmse7_ann[5],Rmse8_ann[5],Rmse9_ann[5]]
            Day_7_ANN=[Rmse7_ann[6],Rmse8_ann[6],Rmse9_ann[6]]
            Day_8_ANN=[Rmse7_ann[7],Rmse8_ann[7],Rmse9_ann[7]]
            Day_9_ANN=[Rmse7_ann[8],Rmse8_ann[8],Rmse9_ann[8]]
            Day_10_ANN=[Rmse7_ann[9],Rmse8_ann[9],Rmse9_ann[9]]
[62]: RF_P=Day_1_rf+Day_2_rf+Day_3_rf+Day_4_rf+Day_5_rf+Day_6_rf+Day_7_rf+Day_8_rf+Day_9_rf+Day_10_r
            XGBoost_P=Day_1_ann+Day_2_ann+Day_3_ann+Day_4_ann+Day_5_ann+Day_6_ann+Day_7_ann+Day_8_ann+Day_
            RF_R=Day_1_RF+Day_2_RF+Day_3_RF+Day_4_RF+Day_5_RF+Day_6_RF+Day_7_RF+Day_8_RF+Day_9_RF+Day_10_F
            XGBoost_R=Day_1_ANN+Day_2_ANN+Day_3_ANN+Day_4_ANN+Day_5_ANN+Day_6_ANN+Day_7_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+Day_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_ANN+DAY_8_
            x1=['5' for i in range(3)]
            x2=['10' for i in range(3)]
            x3=['15' for i in range(3)]
            x4=['20' for i in range(3)]
            x5=['25' for i in range(3)]
            x6=['30' for i in range(3)]
            x7=['35' \text{ for i in range}(3)]
            x8=['40' for i in range(3)]
            x9=['45' \text{ for i in range}(3)]
            x10=['50' for i in range(3)]
            x=x1+x2+x3+x4+x5+x6+x7+x8+x9+x10
[63]: fig = go.Figure()
            fig.add_trace(go.Box(
                     # defining y axis in corresponding
                     # to x-axis
                     y=RF_P,
                     x=x,
```

```
name='RF(Pearson r)',
    marker_color='darkblue',
    showlegend=True
))
fig.add_trace(go.Box(
   y=XGBoost_P,
    x=x,
    name='XGBoost(Pearson r)',
    marker_color='hotpink',
    showlegend=True
))
fig.add_trace(go.Box(
    # defining y axis in corresponding
    # to x-axis
    y=RF_R,
    x=x,
    name='RF(NSRMSE)',
    marker_color='olive',
    showlegend=True
))
fig.add_trace(go.Box(
    y=XGBoost_R,
    x=x.
    name='XGBoost(NSRMSE)',
    marker_color='orangered',
    showlegend=True
))
fig.update_layout(autosize=False,
                   title={'text': "<b>Seasonal Calibration Scheme (CO__
→Data)(Unshuffled)</b>",
        'y':0.83,
        'x':0.5,
        'xanchor': 'center',
        'yanchor': 'top'},
    width=800,
    height=500,
    legend=dict( yanchor="bottom",
    y=1.05,
    orientation="h"
),
    # group together boxes of the different
    # traces for each value of x
```

[63]:



```
[64]: df1=[x for _, x in CO_Data.groupby('Month')]
    data_oct=df1[4]
    #data_oct=data_oct.sample(frac=1)
    data_nov=df1[5]
    #data_nov=data_nov.sample(frac=1)
    data_dec=df1[6]
    #data_dec=data_dec.sample(frac=1)
    data_jan=df1[0]
    #data_jan=data_jan.sample(frac=1)
    data_feb=df1[1]
    #data_feb=data_feb.sample(frac=1)
```

```
data_mar=df1[2]
#data_mar=data_mar.sample(frac=1)
data_apr=data_apr.sample(frac=1)
data_apr=data_apr.sample(frac=1)
data=[data_oct,data_nov,data_dec,data_jan,data_feb,data_mar]
frame1=[data_dec] #, data_nov, data_dec, data_jan, data_feb, data_mar, data_apr]
fall=pd.concat(frame1)
fall.shape
```

[64]: (741, 9)

```
[65]: fall train=fall[:5760]
      fall_train=fall_train.sample(frac=1)
      fall test=fall[6000:]
      Day=[i for i in range(1,21)]
      X_train=fall_train[['Net_
      →Signal', 'Lab1', 'Temp', 'RH', 'Month', 'Day_of_week', 'Day', 'Hour']]
      y_train=fall_train['Ref']
      X test=fall test[['Net]]
      →Signal', 'Lab1', 'Temp', 'RH', 'Month', 'Day_of_week', 'Day', 'Hour']]
      y test=fall test['Ref']
      mean1=np.std(y)
      Rmse7_rf=[]
      RMSE7 rf=[]
      for i in range(1,21):
          regressor.fit(X_train[:288*i].drop(['Lab1'], axis=1), y_train[:288*i])
          pred=regressor.predict(X_test.drop(['Lab1'], axis=1))
          #y_test=y_test
          for i in range(len(y_test)):
              if y_test[i]>3000:
                  y_test[i]=np.mean(y_test)
          mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
          rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
          Rmse7_rf.append(mse)
          RMSE7_rf.append(rmse)
```

```
/Library/Frameworks/Python.framework/Versions/3.8/lib/python3.8/site-packages/
→sklearn/ensemble/_forest.py in predict(self, X)
                check_is_fitted(self)
    782
                # Check data
--> 783
                X = self. validate X predict(X)
    784
    785
                # Assign chunk of trees to jobs
/Library/Frameworks/Python.framework/Versions/3.8/lib/python3.8/site-packages/
→sklearn/ensemble/_forest.py in _validate_X_predict(self, X)
    419
                check_is_fitted(self)
    420
--> 421
                return self.estimators_[0]._validate_X_predict(X,_
422
    423
            @property
/Library/Frameworks/Python.framework/Versions/3.8/lib/python3.8/site-packages/
→sklearn/tree/ classes.py in validate X predict(self, X, check input)
                """Validate X whenever one tries to predict, apply,
→predict proba"""
    387
                if check_input:
                     X = check_array(X, dtype=DTYPE, accept_sparse="csr")
--> 388
    389
                     if issparse(X) and (X.indices.dtype != np.intc or
    390
                                          X.indptr.dtype != np.intc):
/Library/Frameworks/Python.framework/Versions/3.8/lib/python3.8/site-packages/
→sklearn/utils/validation.py in inner f(*args, **kwargs)
     71
                                   FutureWarning)
     72
                kwargs.update({k: arg for k, arg in zip(sig.parameters, args)})
---> 73
                return f(**kwargs)
     74
            return inner f
     75
/Library/Frameworks/Python.framework/Versions/3.8/lib/python3.8/site-packages/
→sklearn/utils/validation.py in check_array(array, accept_sparse, 

→accept_large_sparse, dtype, order, copy, force_all_finite, ensure_2d, 
→allow_nd, ensure_min_samples, ensure_min_features, estimator)
                n samples = num samples(array)
    649
    650
                if n_samples < ensure_min_samples:</pre>
--> 651
                     raise ValueError("Found array with %d sample(s) (shape=%s)
⇒while a"
    652
                                       " minimum of %d is required%s."
    653
                                      % (n_samples, array.shape,__
→ensure_min_samples,
```

```
ValueError: Found array with 0 sample(s) (shape=(0, 7)) while a minimum of 1 is 

→required.
```

```
[]: index=[i for i in range(len(y_test))]
     plt.plot(index[:100],y_test[:100])
     plt.plot(index[:100],pred[:100])
[]: df1=[x for _, x in CO_Data.groupby('Month')]
     data oct=df1[4]
     #data_oct=data_oct.sample(frac=1)
     data nov=df1[5]
     #data_nov=data_nov.sample(frac=1)
     data dec=df1[6]
     #data_dec=data_dec.sample(frac=1)
     data jan=df1[0]
     \#data_jan=data_jan.sample(frac=1)
     data feb=df1[1]
     \#data\_feb=data\_feb.sample(frac=1)
     data_mar=df1[2]
     #data_mar=data_mar.sample(frac=1)
     data_apr=df1[3]
     #data_apr=data_apr.sample(frac=1)
     data=[data_oct,data_nov,data_dec,data_jan,data_feb,data_mar,data_apr]
     {\tt frame2=[data\_dec]\#, data\_nov, data\_dec, data\_jan, data\_feb, data\_mar, data\_apr]}
     fall2=pd.concat(frame2)
     fall2=fall2.sample(frac=1)
[]: from sklearn.model selection import train test split
     X=fall2[['Net Signal','Lab1','Temp','RH','Month','Day_of_week','Day','Hour']]
     y=fall2['Ref']
     \#X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y, test\_size = 0.069)
     mean1=np.std(y)
     Rmse7_rf=[]
     RMSE8_rf=[]
     for i in range(1,21):
         regressor.fit(X[:288*i].drop(['Lab1'], axis=1), y[:288*i])
         pred=regressor.predict(X[6000:].drop(['Lab1'], axis=1))
         y_test=y[6000:]
         for i in range(len(y test)):
             if y_test[i]>3000:
                 y test[i]=np.mean(y test)
         mse=round(np.sqrt(sm.mean_squared_error(y_test, pred))/np.mean(y_test),2)
         rmse=round(np.corrcoef(y_test, pred)[0, 1],2)
         Rmse7_rf.append(mse)
         RMSE8_rf.append(rmse)
```

```
[]: index=[i for i in range(len(y_test))]
    plt.plot(index[:100],y_test[:100])
    plt.plot(index[:100],pred[:100])
[]: fig= plt.figure(figsize=(12,6))
    ax = fig.add_subplot(111)
    #ax.patch.set facecolor('lightblue')
    \#ax.patch.set_alpha(0.3)
    #plt.plot(Day,np.array(Rmse7_rf)+0.6,color='teal',linewidth=2)
    plt.plot(Day,RMSE7_rf,color='salmon',linewidth=2)
    plt.plot(Day,RMSE8_rf,color='teal',linewidth=2)
    plt.legend(['Non-randomized', 'Randomized'], loc = 2, bbox_to_anchor = (0,1))
    plt.xlabel('Days of training',fontsize=18)
    plt.ylabel('Pearson r',fontsize=18)
    plt.title('Random Forest(RF) 6-Month Calibration Scheme using CO_{\sqcup}
     #plt.grid(linestyle='-.',linewidth=0.3)
[]: max(y_test)
[]: max(pred)
[]:
```