

General information

This Colab notebook enables to design sequence of IDPs with target scaling exponents (ν).

The user needs to provide as input a starting sequence and a target value for ν . The starting sequence can be evolved with swap moves or single point mutations.

If you use this notebook, please cite:

- Pesce, F., Bremer, A., Tesei, G., Hopkins, J. B., Grace, C. R., Mittag, T., & Lindorff-Larsen, K. (2023). Design of intrinsically disordered protein variants with diverse structural properties. bioRxiv. DOI: https://doi.org/10.1101/2023.10.22.563461
- Tesei, G., Trolle, A. I., Jonsson, N., Betz, J., Knudsen, F. E., Pesce, F., ... & Lindorff-Larsen, K. (2024). Conformational ensembles of the human intrinsically disordered proteome. *Nature*, 626(8000), 897-904. DOI: https://doi.org/10.1038/s41586-023-07004-5

```
In [1]: #@title <b>Preliminary operations (i)</b>
        import random
        import subprocess
        import os
        import pandas as pd
        pd.options.mode.chained_assignment = None
        import numpy as np
        import itertools
        from joblib import dump, load
        # subprocess.run( 'pip install wget localcider==0.1.18'.split() )
        # subprocess.run('pip uninstall scikit-learn -y'.split())
        # subprocess.run('pip install scikit-learn==1.0.2'.split())
        import wget
        from localcider.sequenceParameters import SequenceParameters
        import matplotlib.pyplot as plt
        # from google.colab import files
        from ipywidgets import IntProgress
        from IPython.display import display
        from IPython.display import clear_output
        import warnings
        warnings.filterwarnings('ignore')
        url = 'https://github.com/KULL-Centre/_2023_Tesei_IDRome/blob/main'
        if os.path.exists('svr_model_nu.joblib') == False:
          wget.download(url+'/svr_models/svr_model_nu.joblib?raw=true')
        if os.path.exists('residues.csv') == False:
          wget.download(url+'/md_simulations/data/residues.csv?raw=true')
```

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def calc_seq_prop(seq,residues,pH=7.):
    seq = list(seq).copy()
   fasta_kappa = np.array(seq.copy())
   N = len(seq)
    r = residues.copy()
   # calculate properties that do not depend on charges
   fK = sum([seq.count(a) for a in ['K']])/N
   fR = sum([seq.count(a) for a in ['R']])/N
   fE = sum([seq.count(a) for a in ['E']])/N
   fD = sum([seq.count(a) for a in ['D']])/N
    faro = sum([seq.count(a) for a in ['W', 'Y', 'F']])/N
   mean_lambda = np.mean(r.loc[seq].lambdas)
    pairs = np.array(list(itertools.combinations(seq,2)))
    pairs_indices = np.array(list(itertools.combinations(range(N),2))
   # calculate sequence separations
   ij_dist = np.diff(pairs_indices,axis=1).flatten().astype(float)
    # calculate lambda sums
   ll = r.lambdas.loc[pairs[:,0]].values+r.lambdas.loc[pairs[:,1]]
   # calculate SHD
    beta = -1
    shd = np.sum(ll*np.power(np.abs(ij_dist),beta))/N
   # fix charges
    r.loc['X'] = r.loc[seq[0]]
    r.loc['X','q'] = r.loc[seq[0],'q'] + 1.
    seq[0] = 'X'
    if r.loc['X','q'] > 0:
        fasta_kappa[0] = 'K'
    else:
        fasta_kappa[0] = 'A'
    r.loc['Z'] = r.loc[seq[-1]]
    r.loc['Z', 'q'] = r.loc[seq[-1], 'q'] - 1.
    seq[-1] = 'Z'
    if r.loc['Z','q'] < 0:</pre>
        fasta_kappa[-1] = 'D'
    else:
        fasta kappa[-1] = 'A'
   Hc = 1/(1+10**(pH-6))
    if Hc < 0.5:
        r.loc['H', 'q'] = 0
        fasta_kappa[np.where(np.array(seq) == 'H')[0]] = 'A'
    elif Hc >= 0.5:
        r.loc['H', 'q'] = 1
        fasta_kappa[np.where(np.array(seq) == 'H')[0]] = 'K'
   # calculate properties that depend on charges
    pairs = np.array(list(itertools.combinations(seq,2)))
    # calculate charge products
   qq = r.q.loc[pairs[:,0]].values*r.q.loc[pairs[:,1]].values
    # calculate SCD
    scd = np.sum(qq*np.sqrt(ij_dist))/N
    SeqOb = SequenceParameters(''.join(fasta_kappa))
    kappa = Seq0b.get_kappa()
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omega_aro = SeqOb.get_kappa_X(grp1=['Y','F','W'])
fcr = r.q.loc[seq].abs().mean()
ncpr = r.q.loc[seq].mean()

return pd.Series(data=[fK,fR,fE,fD,faro,scd,shd,kappa,omega_aro index=['fK','fR','fE','fD','faro','SCD','SHD','kap
```

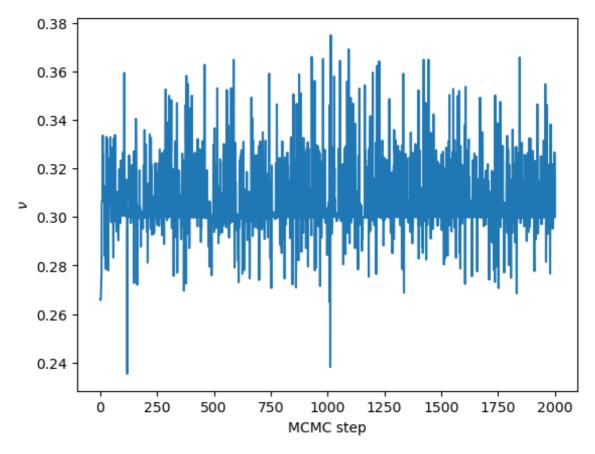
```
In [2]: #@title <b>Preliminary operations (ii)</b>
        class Sequence:
          aa = ['A','C','D','E','F','G','H','I','K','L','M','N','P','Q','R'
          def __init__(self, start_seq=None, length=None, pH=7.):
            assert start_seq is not None or length is not None, "Either ini
            self.pH = pH
            if start_seq is None and length is not None:
              self.sequence = random.choices(self.aa, k=length)
            else:
              self.sequence = list(start_seq)
            self.length = len(self.sequence)
            self.residues = pd.read_csv('residues.csv').set_index('one')
            self.model_nu = load('svr_model_nu.joblib')
          def get sequence(self):
            return ''.join(self.sequence)
          def mutate(self, mode):
            if mode == 'single_point':
              self.sequence[random.randrange(len(self.sequence))] = random.
            if mode == 'swap_full':
              a, b = random.randrange(len(self.sequence)), random.randrange
              self.sequence[b], self.sequence[a] = self.sequence[a], self.s
            if mode == 'swap_expr':
              a, b = random.randrange(1,len(self.sequence)), random.randran
              self.sequence[b], self.sequence[a] = self.sequence[a], self.s
          def featurizer(self):
            self.features = calc_seq_prop(self.sequence,self.residues,self.
          def get_nu(self):
            features_nu = ['SCD','SHD','kappa','FCR','mean_lambda']
            self.featurizer()
            return self.model_nu.predict(self.features.loc[features_nu].val
        class MCMC(Sequence):
          def __init__(self, mutation_mode, nu_target, start_seq=None, leng
            super().__init__(start_seq, length, pH)
            self.memory = pd.DataFrame(columns=['fasta','nu','mc'])
            self.memory.loc[0] = dict(fasta=self.sequence.copy(),
                                nu=self.get_nu(),
                                mc=True)
            self.Features = pd.DataFrame(columns=['fK','fR','fE','fD','faro
            self.Features.loc[0] = self.features.copy()
            self.mode = mutation_mode
            self.nu_target = nu_target
            self.c = c
```

```
def step(self):
            self.mutate(mode = self.mode)
            self.memory.loc[self.memory.index.values[-1]+1] = dict(fasta=se
            self.Features.loc[self.Features.index.values[-1]+1] = self.feat
            lastmc = self.memory[(self.memory['mc']==True)].index[-1]
            L = abs(self.memory.nu[self.memory.index.values[-1]] - self.nu_
            self.memory.mc[self.memory.index.values[-1]] = np.exp(L/-self.c)
            if self.memory.mc[self.memory.index.values[-1]] == False:
              self.sequence = self.memory.fasta[lastmc].copy()
            if len(self.memory) % (self.length*2) == 0:
              self.c = self.c*0.99
In [3]: import sklearn
        #@title <b>Input sequence and parameters</b>
        NAME = "Q9UHJ3_down" #@param {type:"string"}
        SEQUENCE = "VLTKTKYTHYYGKKKNKRIGRPPGGHSNLACALKKASKRRKRKNVFVHKKKRSS
        nu_TARGET = 0.3 #@param {type:"number"}
        MUTATION_TYPE = 'swap_full' # @param ["swap_full", "swap_expr", "si
        CONTROL_PARAMETER = 0.00002 #@param {type:"number"}
        if " " in SEQUENCE:
            SEQUENCE = ''.join(SEQUENCE.split())
        evo = MCMC(start_seq = SEQUENCE, mutation_mode=MUTATION_TYPE, nu_ta
In [4]: #@title <b>Evolve</b>
        STEPS = 2000 #@param {type:"number"}
        f = IntProgress(min=0, max=STEPS, description='Progress:', bar_styl
        display(f)
        for i in range(STEPS):
          evo.step()
          f.value += 1
        clear_output()
        acc rate = sum(evo.memory.mc.values)/len(evo.memory)
        print(f'Acceptance rate: {acc_rate}')
        plt.plot(evo.memory.nu.values)
```

Acceptance rate: 0.2843578210894553

plt.xlabel('MCMC step')
plt.ylabel(r'\$\nu\$')

plt.show()



In [5]: #@title Sequence features
evo.Features

[5]:		fK	fR	fE	fD	faro	SCD	SHI
	0	0.117284	0.08642	0.098765	0.080247	0.037037	-8.619892	3.61491
	1	0.117284	0.08642	0.098765	0.080247	0.037037	-8.619892	3.60868
	2	0.117284	0.08642	0.098765	0.080247	0.037037	-8.619892	3.61491
	3	0.117284	0.08642	0.098765	0.080247	0.037037	-8.619892	3.614976
	4	0.117284	0.08642	0.098765	0.080247	0.037037	-8.581543	3.613060
	•••	•••			•••	•••		
	1996	0.117284	0.08642	0.098765	0.080247	0.037037	-8.498132	3.61190:
	1997	0.117284	0.08642	0.098765	0.080247	0.037037	-8.498132	3.616376
	1998	0.117284	0.08642	0.098765	0.080247	0.037037	-7.563136	3.61169
	1999	0.117284	0.08642	0.098765	0.080247	0.037037	-7.711587	3.611488
	2000	0.117284	0.08642	0.098765	0.080247	0.037037	-8.498132	3.611944

2001 rows × 12 columns

Out

```
In []: #@title <b>Download results</b>
out = evo.Features.copy()
out['fasta'] = [''.join(i) for i in evo.memory.fasta.values]
out['nu'] = evo.memory.nu
```

```
out.to_csv(f'{NAME}_designs.csv')
# files.download(f'{NAME}_designs.csv')
```