I4 Weak stabilities by consecutive intersect

April 1, 2022

1 Purpose:

As we have seen for I_4 we have $|A_i \cap A_j| = 0$ for $j \neq i+1, i-1$. So by the global condition we have $|A_i \cap A_{i+1}| = 1$ (all other intersects are empty). Using that $|A_i \cap A_{i+1}| = 1$, we provide a procedure to obtain all weak stabilities for I_4 .

1.1 Procedure to obtain assignments

With the trees labeled as in overleaf and assignment $\underline{0}$ on T_1 , we determine possible \underline{d}_2 on T_2 by permuting through possible vertex values that will allow for $|\sigma(T_1,\underline{0}) \cap \sigma(T_2,\underline{d}_2)| = 1$. We then determine \underline{d}_3 on T_3 by iterating over valid \underline{d}_2 terms and considering $|\sigma(T_2,\underline{d}_2) \cap \sigma(T_3,\underline{d}_3)| = 1$, similarly for I_4 .

This however does not give all stabilities obtained by ϕ (see overleaf document). So I propose we also have to go the anti-clockwise too. That is consider T_1, T_4 to determine \underline{d}_4 , then T_4, T_3 , and then T_3, T_2 . These are the only ways to go on I_4 so we are done.

1.2 Notation: Consistency with overleaf.

For I_4 we label this as in overleaf for a list we have $[v_1, v_2, v_3, v_4]$ with assignments respectively $[d_1, d_2, d_3, d_4]$ and similar for items of $\sigma(\underline{d})$. Such labeling will allow us to compare to stability conditions determined by ϕ . Let $A_i := \sigma(T_i, \underline{d}_i)$.

```
[177]: from IPython.display import Image
[178]: fig = Image(filename=('I4_weak_stab\I4_notation.png'), width=700, height=300)
    fig
[178]:
```

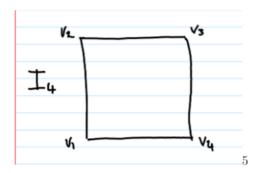


Figure 1: Labelled I_4

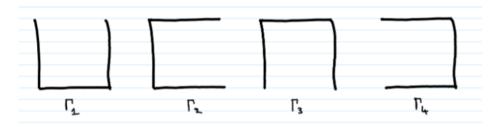


Figure 2: Ordered spanning trees of I_4

1.3 Punchline

This procedure gives the following weak stabilties, for the clockwise we have the following:

The output of this procedure anticlockwise gives the following:

```
[196]: fig = Image(filename=('I4_weak_stab\clockwise2.png'), width=700, height=400) fig
```

[196]:

Analysis Auti- Clochwise.

These values can be seen from the ϕ stabiltiy conditions we previously calculated.

1.4 Load general data

```
[87]: from sympy import *
      from collections import OrderedDict
      init_printing()
[88]: #Symbols
      d11, d12, d13, d14 = symbols("d11, d12, d13,d14", integer=True)
[89]: #Assignments:
      d1 = Matrix([0,0,0,0])
      # For sigma data
      v11= Matrix([0,1,0,0]) #For T_1
      v12= Matrix([0,0,1,0])
      v21= Matrix([0,0,1,0]) #For T_2
      v22= Matrix([0,0,0,1])
      v31= Matrix([0,0,0,1]) #For T_3
      v32= Matrix([1,0,0,0])
      v41= Matrix([1,0,0,0]) #For T_4
      v42= Matrix([0,1,0,0])
      sig d1=FiniteSet(d1+v11,d1+v12)
      sig=sig_d1 #Change for Case 2) and 3).
```

2 Clockwise

2.1 Case 1: T1,T2: Obtaining d2 list

As $|\sigma(T_1,\underline{0}) \cap \sigma(T_2,\underline{d_2})|=1$, the range for $d_2(v_i)$ are given by the following. Need choices for $a=d_2(v_1)$, $b=d_2(v_2)$, $c=d_2(v_3)$, $d=d_2(v_4)$ values for d_2 from $\sigma(T_1,\underline{0})$.

 $\{0\}$ $\{0, 1\}$ $\{0, 1, -1\}$ $\{0, -1\}$

Now permute through possible values on vertices to get valid d_2 , to be checked for size later.

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2.2 Case 2: T2,T3: Obtaining d3 list

```
[94]: # Need choices for a,b,c,d values for d3 from sig_d2 over possible d2 terms.

d3_lst=[] #Total list of assignments d3
```

```
for d2 in d2_lst:
    d2=Matrix(d2)
    sig_d2=FiniteSet(d2+v21,d2+v22)
    sig=sig_d2
    a_lst,b_lst,c_lst,d_lst=[],[],[],[]
    for i in sig:
        a_lst.append(i[0])
        b_lst.append(i[1])
        c_lst.append(i[2])
        d_lst.append(i[3])
    a set,b_set,c_set,d set=set(a lst),set(b lst),set(c lst),set(d lst)
    ## As d3 on T3 we include the values of the shift of sigma [-1,0,0,-1].
 \rightarrow (where we have +1 on T_3 by chip adding).
    d_set_m={x-1 for x in d_set}
    a_set_m={x-1 for x in a_set}
    d_set=d_set.union(d_set_m)
    a_set=a_set.union(a_set_m)
    # Now permute through possible values on vertices.
    partial_d3_lst=[] #Recording d3 assignments for this d2
    for a in a_set:
        for b in b_set:
            for c in c_set:
                for d in d_set:
                    d3=[a,b,c,d]
                    partial_d3_lst.append(d3)
    d3_lst=d3_lst+partial_d3_lst # Adding to total d3 list.
print(len(d3_lst))
```

2.3 Case 3: T3,T4: Obtaining d4 list

```
[95]: # Need choices for a,b,c,d values for d4.
      d4_lst=[]
      for d3 in d3_lst:
          d3=Matrix(d3)
          sig_d3=FiniteSet(d3+v31,d3+v32)
          sig=sig_d3
          a_lst,b_lst,c_lst,d_lst=[],[],[],[]
          for i in sig:
              a_lst.append(i[0])
              b_lst.append(i[1])
              c_lst.append(i[2])
              d_lst.append(i[3])
          a_set,b_set,c_set,d_set=set(a_lst),set(b_lst),set(c_lst),set(d_lst)
          ## As d3 on T3 we include the values of the shift [-1,-1,0,0]. (where we
       \rightarrow have +1 on T_4 by chip adding).
          a_set_m={x-1 for x in a_set}
          b_set_m={x-1 for x in b_set}
          a_set=a_set.union(a_set_m)
          b_set=b_set.union(b_set_m)
          # Now permute through possible values on vertices for this d3
          partial_d4_lst=[]
          for a in a_set:
              for b in b_set:
                  for c in c_set:
                      for d in d_set:
                          d4=[a,b,c,d]
                          partial_d4_lst.append(d4)
          d4_lst=d4_lst+partial_d4_lst
      print(len(d4_lst))
```

2.4 Assignments giving $|\sigma^{(d1,d2,d3,d4)}(I_4)| = 4$

```
[96]: # Get assignments, by taking all combinations.

ass_lst=[]
for d2 in d2_lst:
    for d3 in d3_lst:
        for d4 in d4_lst:
            ass=[d1,d2,d3,d4]
            ass_lst.append(ass)

print(len(ass_lst)) #2985984
```

2985984

```
[97]: #ass_lst will have repeats of assignments: we remove them.

#Want to kill repeats of list of lists.
lst =ass_lst
t_lst=[tuple([tuple(i) for i in x]) for x in lst]
no_repeats=list(OrderedDict.fromkeys(t_lst))#
no_reps_lst=[[list(i) for i in x] for x in no_repeats]

ass_lst=no_reps_lst
print(len(ass_lst)) #184320
```

```
[116]: #Checking for which assignments we get sigma of size 4

def checker(start,end):

    v11= Matrix([0,1,0,0]) #For T_1
    v12= Matrix([0,0,1,0]) #For T_2
    v21= Matrix([0,0,0,1]) #For T_2
    v22= Matrix([0,0,0,1]) #For T_3
    v32= Matrix([1,0,0,0])

    v41= Matrix([1,0,0,0]) #For T_4
    v42= Matrix([0,1,0,0])

    good_assigns=[]
    for ass in ass_lst[start:end]:
        d1=Matrix(ass[0])
        d2=Matrix(ass[1])
```

```
[118]: # Breaking into cases:

good_ass=[]

for i in range(0,184320,10000):
    if i ==180000:
        check=checker(i,i+4320)
        print(f"Check #{i}-{i+4320}: Number of good assignments is: {check[0]}")
        good_ass=good_ass+check[1]
        break
    check=checker(i,i+10000)
    print(f"Check #{i}-{i+10000}: Number of good assignments is: {check[0]}")
        good_ass=good_ass+check[1]
```

```
Check #0-10000: Number of good assignments is: 2
Check #10000-20000: Number of good assignments is: 0
Check #20000-30000: Number of good assignments is: 0
Check #30000-40000: Number of good assignments is: 0
Check #40000-50000: Number of good assignments is: 0
Check #50000-60000: Number of good assignments is: 0
Check #60000-70000: Number of good assignments is: 0
Check #70000-80000: Number of good assignments is: 0
Check #80000-90000: Number of good assignments is: 0
Check #90000-100000: Number of good assignments is: 0
Check #100000-110000: Number of good assignments is: 1
Check #110000-120000: Number of good assignments is: 1
Check #120000-130000: Number of good assignments is: 0
Check #130000-140000: Number of good assignments is: 0
Check #140000-150000: Number of good assignments is: 0
Check #150000-160000: Number of good assignments is: 0
Check #160000-170000: Number of good assignments is: 0
```

```
Check #170000-180000: Number of good assignments is: 0 Check #180000-184320: Number of good assignments is: 0
```

2.5 Analysis

We now compare this to the weak stabilities in overleaf for I_4 .

```
[119]: print(f"The total length of good_assignments is: {len(good_ass)}")

for i in good_ass:
    print(i)
```

```
The total length of good_assignments is: 4

[[0, 0, 0, 0], [0, 0, 0], [0, 0, 0, 0], [0, 0, 0, 0]]

[[0, 0, 0, 0], [0, 0, 0, 0], [-1, 1, 0, 0], [-1, 0, 0, 1]]

[[0, 0, 0, 0], [0, 1, 0, -1], [0, 0, 1, -1], [0, 0, 1, -1]]

[[0, 0, 0, 0], [0, 1, 0, -1], [-1, 1, 1, -1], [-1, 0, 1, 0]]
```

These are weak stabilities labeled 2,3,4,5 in overleaf. Where are 1,6? Note for these weak stabilities $d_2[0]=0$ which is from T_1,T_2 and the construction of d_2 .

```
[120]: def check_sig(ass):
               d1=Matrix(ass[0])
               d2=Matrix(ass[1])
               d3=Matrix(ass[2])
               d4=Matrix(ass[3])
               #Sigma terms/presentations
               d1_pres=FiniteSet(d1+v11,d1+v12)
               d2_pres=FiniteSet(d2+v21,d2+v22)
               d3_pres=FiniteSet(d3+v31,d3+v32)
               d4_pres=FiniteSet(d4+v41,d4+v42)
               total_sig=d1_pres.union(d2_pres).union(d3_pres).union(d4_pres) #does_u
        \rightarrow work.
                  if len(total_sig)==4:
       #
                      qood_assigns.append(ass)
               return [len(total_sig),total_sig]
```

For each assignment we can compare the sigma too.

```
[128]: for i in [good_ass[1]]:
    print("Assignment:",i)

    print(f"\n Sigma:")

    for j in check_sig(i)[1]:
        print(j)
```

```
# print("sigma:",check_sig(i)[1])
```

```
Assignment: [[0, 0, 0, 0], [0, 0, 0, 0], [-1, 1, 0, 0], [-1, 0, 0, 1]]

Sigma:
Matrix([[-1], [1], [0], [1]])
Matrix([[0], [0], [0], [1]])
Matrix([[0], [0], [1], [0]])
Matrix([[0], [1], [0], [0]])
```

3 Anti-Clockwise

We repeat the procedure above but this time do T_1, T_4 , then T_4, T_3 , then T_3, T_2 . This will give a wider range of d_2 on d_2 but smaller range of d_4 on d_4 .

3.1 Load data

```
[140]: #Assignments:
    d1 = Matrix([0,0,0,0])
    sig_d1=FiniteSet(d1+v11,d1+v12)
    sig=sig_d1
```

3.2 Case 1: T1,T4: Obtaining d4 list

```
12 {0, -1} {0, 1, -1} {0, 1} {0}
```

3.3 Case 2: T4,T3: Obtaining d3 list

```
[143]: # Need choices for a,b,c,d values for d3 from sig_d4 over possible d4 terms.
       d3_lst=[] #Total list of assignments d3
       for d4 in d4_lst:
           d4=Matrix(d4)
           sig_d4=FiniteSet(d4+v41,d4+v42)
           sig=sig_d4
           a_lst,b_lst,c_lst,d_lst=[],[],[],[]
           for i in sig:
               a_lst.append(i[0])
               b_lst.append(i[1])
               c_lst.append(i[2])
               d_lst.append(i[3])
           a_set,b_set,c_set,d_set=set(a_lst),set(b_lst),set(c_lst),set(d_lst)
           ## As d3 on T3 we include the values of the shift of sigma [-1,0,0,-1].
        \rightarrow (where we have +1 on T_3 by chip adding).
           d_set_m={x-1 for x in d_set}
           a_set_m={x-1 for x in a_set}
           d_set=d_set.union(d_set_m)
```

```
a_set=a_set.union(a_set_m)

# Now permute through possible values on vertices.

partial_d3_lst=[] #Recording d3 assignments for this d4

for a in a_set:
    for b in b_set:
        for c in c_set:
            for d in d_set:
                d3=[a,b,c,d]
                partial_d3_lst.append(d3)

d3_lst=d3_lst+partial_d3_lst # Adding to total d3 list.

print(len(d3_lst))
```

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3.4 Case 3: T3,T2: Obtaining d2 list

```
[144]: # Need choices for a,b,c,d values for d2.
       d2_lst=[]
       for d3 in d3_lst:
           d3=Matrix(d3)
           sig_d3=FiniteSet(d3+v31,d3+v32)
           sig=sig_d3
           a_lst,b_lst,c_lst,d_lst=[],[],[],[]
           for i in sig:
               a_lst.append(i[0])
               b_lst.append(i[1])
               c_lst.append(i[2])
               d_lst.append(i[3])
           a_set,b_set,c_set,d_set=set(a_lst),set(b_lst),set(c_lst),set(d_lst)
           ## As d2 on T2 we include the values of the shift of sigma [0,0,-1,-1]_{\sqcup}
        \hookrightarrow (where we have +1 on T_2 by chip adding).
           c_set_m={x-1 for x in c_set}
           d_set_m={x-1 for x in d_set}
```

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3.5 Assignments giving $|\sigma^{(d1,d2,d3,d4)}(I_4)| = 4$

```
[145]: # Get assignments, by taking all combinations.

ass_lst=[]
for d2 in d2_lst:
    for d3 in d3_lst:
        for d4 in d4_lst:
        ass=[d1,d2,d3,d4]
        ass_lst.append(ass)
print(len(ass_lst)) #2985984
```

2985984

```
[146]: #ass_lst will have repeats of assignments: we remove them.

#Want to kill repeats of list of lists.
lst =ass_lst
    t_lst=[tuple([tuple(i) for i in x]) for x in lst]
    no_repeats=list(OrderedDict.fromkeys(t_lst))#
    no_reps_lst=[[list(i) for i in x] for x in no_repeats]

ass_lst=no_reps_lst
    print(len(ass_lst)) #184320
```

```
[147]: # Breaking into cases:
```

```
good_ass=[]
       for i in range(0,184320,10000):
           if i ==180000:
               check=checker(i,i+4320)
               print(f"Check #{i}-{i+4320}: Number of good assignments is: {check[0]}")
               good_ass=good_ass+check[1]
              break
           check=checker(i,i+10000)
           print(f"Check #{i}-{i+10000}: Number of good assignments is: {check[0]}")
           good ass=good ass+check[1]
      Check #0-10000: Number of good assignments is: 2
      Check #10000-20000: Number of good assignments is: 1
      Check #20000-30000: Number of good assignments is: 0
      Check #30000-40000: Number of good assignments is: 0
      Check #40000-50000: Number of good assignments is: 1
      Check #50000-60000: Number of good assignments is: 0
      Check #60000-70000: Number of good assignments is: 0
      Check #70000-80000: Number of good assignments is: 0
      Check #80000-90000: Number of good assignments is: 0
      Check #90000-100000: Number of good assignments is: 0
      Check #100000-110000: Number of good assignments is: 0
      Check #110000-120000: Number of good assignments is: 0
      Check #120000-130000: Number of good assignments is: 0
      Check #130000-140000: Number of good assignments is: 0
      Check #140000-150000: Number of good assignments is: 0
      Check #150000-160000: Number of good assignments is: 0
      Check #160000-170000: Number of good assignments is: 0
      Check #170000-180000: Number of good assignments is: 0
      Check #180000-184320: Number of good assignments is: 0
      3.6 Analysis
[148]: print(f"The total length of good_assignments is: {len(good_ass)}")
       for i in good_ass:
           print(i)
      The total length of good_assignments is: 4
      [[0, 0, 0, 0], [0, 0, 0, 0], [0, 0, 0, 0], [0, 0, 0, 0]]
      [[0, 0, 0, 0], [1, 0, 0, -1], [0, 0, 1, -1], [0, 0, 0, 0]]
      [[0, 0, 0, 0], [0, 1, 0, -1], [-1, 1, 1, -1], [-1, 0, 1, 0]]
```

[[0, 0, 0, 0], [-1, 1, 0, 0], [-1, 1, 0, 0], [-1, 0, 1, 0]]