

QuartPAC: Identifying mutational clusters while utilizing protein quaternary structural data

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Abstract

The **QuartPAC** package is designed to identify mutated amino acid hotspots while accounting for protein quaternary structure. It is meant to work in conjunction with the **iPAC** [Ryslik and Zhao, 2012b], **GraphPAC** [Ryslik and Zhao, 2012a] and **SpacePAC** [Ryslik and Zhao, 2013] packages already available through Bioconductor. Specifically, the package takes as input the quaternary protein structure as well as the mutational data for each subunit of the assembly. It then maps the mutational data onto the protein and performs the algorithms described in **iPAC**, **GraphPAC** and **SpacePAC** to report the statistically significant clusters. By integrating the quaternary structure, **QuartPAC** may identify additional clusters that only become apparent when the different protein subunits are considered together.

1 Introduction

Recent advances in oncogenic pharmacology [Croce, 2008] have led to the creation of a variety of methods that attempt to identify mutational hotspots as these hotspots are often indicative of driver mutations [Wagner, 2007, Zhou et al., 2008, Ye et al., 2010]. Three recent methods, **iPAC**, **GraphPAC** and **SpacePAC** provide approaches to identify such hotspots while accounting for protein tertiary structure. While it has been shown that these mutations provide an improvement over linear clustering methods, [Ryslik et al., 2013, 2014b,a], they nevertheless consider only tertiary structure. **QuartPAC**, preprocesses

the entire assembly structure in order to be able to accurately run these approaches on the quaternary protein unit. This allows for the identification of additional mutational clusters that may otherwise be missed if only one protein subunit is considered at a time.

In order to run **QuartPAC**, four sources of data are required:

- The amino acid sequence of the protein which is obtained from the UniProt database (uniprot.org in FASTA format).
- The protein tertiary subunit information which is obtained from the .pdb file from PDB.org
- The quaternary structural information for the entire assembly which is obtained from the .pdb1 file from PDB.org
- The somatic mutation data which is obtained from the Catalogue of Somatic Mutations in Cancer (<http://cancer.sanger.ac.uk/cancergenome/projects/cosmic/>).

In order to map the mutations onto the protein quaternary structure, an alignment must be performed. For each uniprot within the assembly, mutational data must be provided. The data is in the format of $m \times n$ matrices for every subunit. A "1" in the (i, j) element indicates that residue j for individual i has a mutation while a "0" indicates no mutation. To be compatible with this software, please ensure that your mutation matrices have R column headings of $V1, V2, \dots, Vn$. Only missense mutations are currently supported, indels in the amino acid sequence are not. Sample mutational data are included in this package as textfiles in the *extdata* folder.

It is worth noting that there does not exist any one individual source to obtain mutational data. One common resource is the COSMIC database <http://cancer.sanger.ac.uk/cancergenome/projects/cosmic/>. The easiest way to obtain mutational data for many proteins is to load the COSMIC database on a local sql server and then query the database for the protein of interest. It is important to restrict your query to whole gene screens or whole genome studies to prevent specific mutations from being selectively chosen (and thus violating the uniformity assumption that **iPAC**, **GraphPAC**, and **SpacePAC** rely upon).

Should you find a bug, or wish to contribute to the code base, please contact the author.

2 Identifying Clusters and Viewing the Remapping

The general principle of **QuartPAC** is that we preprocess the data into a format that can be recognized by **iPAC**, **GraphPAC** and **SpacePAC**. Most of this is automated and all that is needed is to point the algorithm to the mutational

and structural data. **QuartPAC** will then reorganize the data, execute the cluster finding algorithms and report the results. The clusters are reported by serial number. As each serial number is unique in the assembly, the user can then map each serial number to the exact atom of interest in the structure.

Below we run the algorithm with no multiple comparison adjustment. We do this to ensure that some clusters are found for each method. We also note that for **iPAC** and **GraphPAC**, if a multiple comparison adjustment is used and no clusters are found significant, the methods will show a null value. For **SpacePAC**, as there is no multiple comparison adjustment needed, the most significant clusters are always shown, regardless of the p-value. This behavior follows the functionality of the previous three packages, so users familiar with the tertiary algorithms will find the results directly comparable.

For more information on the output, please see the **iPAC**, **GraphPAC**, and **SpacePAC** packages as the output is similar. The main difference is that the amino acid numbers now refer to the serial numbers within the *.pdb1 file.

Code Example 1: Running QuartPAC.

```
> library(QuartPAC)
> #read the mutational data
> mutation_files <- list(
+ system.file("extdata", "HFE_Q30201_MutationOutput.txt", package = "QuartPAC"),
+ system.file("extdata", "B2M_P61769_MutationOutput.txt", package = "QuartPAC")
+ )
> uniprots <- list("Q30201", "P61769")
> mutation.data <- getMutations(mutation_files = mutation_files, uniprots = uniprots)
> #read the pdb file
> pdb.location <- "https://files.rcsb.org/view/1A6Z.pdb"
> assembly.location <- "https://files.rcsb.org/download/1A6Z.pdb1"
> structural.data <- makeAlignedSuperStructure(pdb.location, assembly.location)
> #Perform Analysis
> #We use a very high alpha level here with no multiple comparison adjustment
> #to make sure that each method provides shows a result.
> #Lower alpha cut offs are typically used.
> quart_results <- quartCluster(mutation.data, structural.data, perform.ipac = "Y",
+                               perform.graphpac = "Y", perform.spacepac = "Y",
+                               create.map = "Y", alpha = .3, MultComp = "None",
+                               Graph.Title = "MDS Mapping to 1D Space",
+                               radii.vector = c(1:3))
```

We observe that the MDS remapping plot provided by **QuartPAC** is done automatically if the *create.map* parameter is set to "Y". The plot is shown in Figure 1 below.

For the **GraphPAC** approach, the linear "Jump Plot" (see the **GraphPAC** package for more details and interpretation) has been implemented and is shown in Figure 2 below. Feel free to contact the author if you want to assist in porting other graphing functionality.

With regards to **SpacePAC**, as there is no remapping from 3D to 1D space, a plotting option that shows the protein in its folded state is presented in Section

4.

MDS Mapping to 1D Space

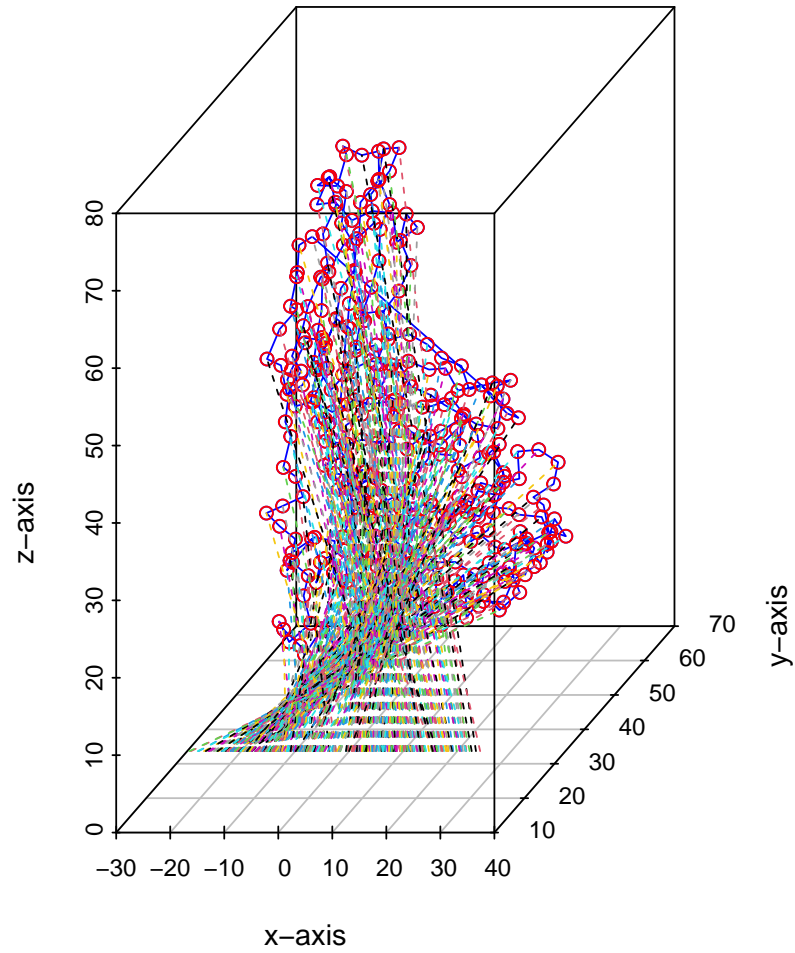


Figure 1: Remapping performed by **iPAC**.

Code Example 2: Plotting the GraphPAC candidate path.

```
> Plot.Protein.Linear(quart_results$graphpac$candidate.path, colCount = 10,  
+                      title = "Protein Reordering to 1D Space via GraphPAC")
```

Protein Reordering to 1D Space via GraphPAC

2	13	19	29	35	43	53	65	73	84
92	96	101	107	116	121	129	137	141	149
155	163	174	183	188	196	200	212	219	227
235	244	252	263	270	281	293	301	311	316
322	333	344	351	360	367	378	385	392	406
411	416	421	426	431	437	443	452	460	474
479	488	496	501	510	516	524	533	537	551
559	569	577	588	595	602	610	621	635	642
650	658	667	675	685	693	703	709	718	727
733	743	750	758	767	774	782	790	794	800
809	817	826	831	839	847	853	860	869	873
885	899	908	920	924	936	944	948	957	965
975	983	992	1003	1009	1016	1024	1031	1039	1047
1061	1072	1077	1082	1091	1098	1109	1114	1128	1135
1142	1151	1159	1168	1182	1191	1202	1212	1221	1229
1240	1245	1256	1265	1273	1284	1289	1301	1309	1318
1329	1337	1343	1350	1355	1364	1372	1381	1390	1398
1406	1415	1423	1427	1432	1436	1443	1451	1459	1468
1477	1484	1491	1498	1506	1513	1522	1529	1536	1546
1556	1563	1570	1576	1582	1589	1596	1603	1611	1622
1628	1639	1644	1652	1660	1672	1684	1691	1700	1708
1716	1723	1731	1740	1754	1762	1771	1779	1788	1797
1804	1812	1820	1825	1834	1843	1854	1863	1870	1879
1887	1894	1902	1909	1917	1921	1929	1933	1940	1952
1961	1965	1979	1987	1994	2002	2007	2014	2021	2028
2032	2041	2050	2059	2070	2082	2089	2095	2104	2111
2120	2130	2137	2141	2149	2157	2166	2173	2181	2189
2196	2204	2220	2228	2237	2248	2255	2262	2271	2279
2288	2295	2307	2313	2324	2334	2341	2346	2355	2363
2367	2376	2382	2390	2401	2409	2417	2423	2435	2442
2448	2452	2463	2473	2480	2486	2494	2502	2511	2518
2526	2534	2542	2551	2559	2563	2572	2583	2591	2600
2605	2612	2621	2631	2637	2645	2653	2659	2670	2676
2685	2693	2707	2713	2724	2736	2744	2752	2764	2776
2783	2792	2803	2810	2817	2824	2833	2838	2846	2855
2867	2872	2878	2889	2896	2904	2914	2921	2928	2936
2942	2951	2958	2967	2975	2982	2991	3005	3013	3024
3032									

Figure 2: Remapping performed by GraphPAC.

3 Using the Output

Now that we have the results, suppose that we wanted to visualize what the clusters are. For example, we see that the first cluster under the **SpacePAC** method for the optimal combination has two spheres. One sphere is centered at the atom with serial number 1265 and one sphere is centered at the atom with serial number 367.

To see where this matches we can query the *structural.data* list.

Code Example 3: Finding the residue of interest using the SpacePAC method.

```
> #look at the results for the optimal sphere combinations under the SpacePAC approach
> #For clarity we only look at columns 3 - 8 which show the sphere centers.
> quart_results$spacepac$optimal.sphere[,3:8]
```

	Start	End	Positions	MutsCount	Z.Score	Within.Range
46	367	367	367	1	-0.2763147	367
154	1265	1265	1265	1	-0.2763147	1265
267	2166	2166	2166	1	-0.2763147	2166
282	2295	2295	2295	1	-0.2763147	2295
295	2401	2401	2401	1	-0.2763147	2401
318	2583	2583	2583	1	-0.2763147	2583
328	2659	2659	2659	1	-0.2763147	2659
349	2846	2846	2846	1	-0.2763147	2846

```
> #Find the atom with serial number 1265
> required.row <- which(structural.data$aligned_structure$serial == 1265)
> #show the information for that atom
> structural.data$aligned_structure[required.row,]
```

	recordName	serial	atom	altLoc	resName	chainID	resSeq	iCode	xCoord	yCoord
	<char>	<int>	<char>	<char>	<char>	<char>	<int>	<char>	<num>	<num>
1:	ATOM	1265	CA		ASN	A	157		5.743	52.485

	zCoord	occupancy	tempFactor	element	charge	UNP	dbref	protomer	absPos
	<num>	<num>	<num>	<char>	<char>	<fctr>	<fctr>	<int>	<int>
1:	13.919	1	42.28	C		Q30201	1	1	154

```
canonical_pos
<num>
1: 179
>
```

Similarly, suppose you wanted to look at the **iPAC** results. The first cluster goes from serial 2583 and ends at 2846. To get all the residue information for that block, we can do the following:

Code Example 4: Finding the residue of interest using the iPAC method.

```
> #look at the results for the first cluster shown by the ipac method
> quart_results$ipac
```

	AA_in_Cluster	serial_start	serial_end	number	p_value
V254	32	2583	2846	3	0.03093808
V172	22	2659	2846	2	0.04285346
V274	55	2401	2846	4	0.05244236
V254	52	2166	2583	4	0.07721776
V274	14	2295	2401	2	0.08306777
V254	37	2295	2583	3	0.09028823
V172	62	2166	2659	5	0.10970551
V274	29	2166	2401	3	0.12480796
V180	304	367	2846	8	0.12630378
V172	283	367	2659	7	0.16003397
V172	47	2295	2659	4	0.16898693
V278	68	2295	2846	5	0.19414336
V303	83	2166	2846	6	0.24825087
V180	222	367	2166	3	0.25497859

```
> #Find the atoms with serial numbers within the range of 2583 to 2846
> required.rows <- which(structural.data$aligned_structure$serial %in% (2583:2846))
> #show the information for those atoms
> structural.data$aligned_structure[required.rows,]
```

	recordName	serial	atom	altLoc	resName	chainID	resSeq	iCode	xCoord	yCoord
	<char>	<int>	<char>	<char>	<char>	<char>	<int>	<char>	<num>	<num>
1:	ATOM	2583	CA		ILE	B	46		12.885	19.924
2:	ATOM	2591	CA		GLU	B	47		12.538	18.636
3:	ATOM	2600	CA		LYS	B	48		9.370	18.878
4:	ATOM	2605	CA		VAL	B	49		9.120	22.638
5:	ATOM	2612	CA		GLU	B	50		7.189	24.889
6:	ATOM	2621	CA		HIS	B	51		7.867	28.493
7:	ATOM	2631	CA		SER	B	52		5.976	31.273
8:	ATOM	2637	CA		ASP	B	53		7.479	32.937
9:	ATOM	2645	CA		LEU	B	54		10.283	35.432
10:	ATOM	2653	CA		SER	B	55		8.837	38.923
11:	ATOM	2659	CA		PHE	B	56		9.783	42.469
12:	ATOM	2670	CA		SER	B	57		8.257	45.640
13:	ATOM	2676	CA		LYS	B	58		8.243	49.331
14:	ATOM	2685	CA		ASP	B	59		11.745	49.858
15:	ATOM	2693	CA		TRP	B	60		13.026	47.034
16:	ATOM	2707	CA		SER	B	61		13.661	44.800
17:	ATOM	2713	CA		PHE	B	62		12.750	41.139
18:	ATOM	2724	CA		TYR	B	63		10.369	38.996
19:	ATOM	2736	CA		LEU	B	64		9.841	35.241
20:	ATOM	2744	CA		LEU	B	65		7.561	33.020
21:	ATOM	2752	CA		TYR	B	66		8.522	29.543
22:	ATOM	2764	CA		TYR	B	67		5.960	27.185
23:	ATOM	2776	CA		THR	B	68		5.265	23.560
24:	ATOM	2783	CA		GLU	B	69		2.275	21.616
25:	ATOM	2792	CA		PHE	B	70		3.018	20.246
26:	ATOM	2803	CA		THR	B	71		1.565	19.125
27:	ATOM	2810	CA		PRO	B	72		3.166	20.559
28:	ATOM	2817	CA		THR	B	73		3.655	18.871

29:	ATOM	2824	CA	GLU	B	74	5.166	19.946
30:	ATOM	2833	CA	LYS	B	75	8.589	18.374
31:	ATOM	2838	CA	ASP	B	76	8.895	19.872
32:	ATOM	2846	CA	GLU	B	77	10.881	23.063

recordName	serial	atom	altLoc	resName	chainID	resSeq	iCode	xCoord	yCoord
zCoord	occupancy	tempFactor	element	charge	UNP	dbref	protomer	absPos	
<num>	<num>	<num>	<char>	<char>	<fctr>	<fctr>	<int>	<int>	
1:	45.964	1	59.61	C	P61769	2	1	318	
2:	42.427	1	75.66	C	P61769	2	1	319	
3:	40.348	1	82.60	C	P61769	2	1	320	
4:	40.986	1	72.00	C	P61769	2	1	321	
5:	38.587	1	65.40	C	P61769	2	1	322	
6:	37.552	1	59.43	C	P61769	2	1	323	
7:	35.791	1	54.96	C	P61769	2	1	324	
8:	32.724	1	54.99	C	P61769	2	1	325	
9:	33.143	1	43.17	C	P61769	2	1	326	
10:	32.873	1	43.06	C	P61769	2	1	327	
11:	33.932	1	36.56	C	P61769	2	1	328	
12:	35.371	1	40.60	C	P61769	2	1	329	
13:	34.385	1	42.56	C	P61769	2	1	330	
14:	35.795	1	41.40	C	P61769	2	1	331	
15:	33.576	1	34.49	C	P61769	2	1	332	
16:	36.581	1	35.68	C	P61769	2	1	333	
17:	36.218	1	33.38	C	P61769	2	1	334	
18:	38.234	1	42.58	C	P61769	2	1	335	
19:	38.323	1	36.62	C	P61769	2	1	336	
20:	40.370	1	38.86	C	P61769	2	1	337	
21:	41.587	1	49.21	C	P61769	2	1	338	
22:	43.115	1	54.28	C	P61769	2	1	339	
23:	44.155	1	59.26	C	P61769	2	1	340	
24:	45.397	1	66.25	C	P61769	2	1	341	
25:	48.852	1	59.77	C	P61769	2	1	342	
26:	52.163	1	67.36	C	P61769	2	1	343	
27:	55.294	1	68.86	C	P61769	2	1	344	
28:	58.669	1	73.63	C	P61769	2	1	345	
29:	61.984	1	84.25	C	P61769	2	1	346	
30:	61.405	1	87.62	C	P61769	2	1	347	
31:	57.911	1	77.81	C	P61769	2	1	348	
32:	57.189	1	68.89	C	P61769	2	1	349	

zCoord	occupancy	tempFactor	element	charge	UNP	dbref	protomer	absPos
canonical_pos								

	<num>
1:	66
2:	67
3:	68
4:	69
5:	70
6:	71
7:	72
8:	73

```

9:          74
10:         75
11:         76
12:         77
13:         78
14:         79
15:         80
16:         81
17:         82
18:         83
19:         84
20:         85
21:         86
22:         87
23:         88
24:         89
25:         90
26:         91
27:         92
28:         93
29:         94
30:         95
31:         96
32:         97
    canonical_pos

```

As the **GraphPAC** results are in the same format as the **iPAC** results, the approach for identifying clusters in those atoms is identical as in the example above.

4 Visualizing the Results

Once you have the serial numbers of interest, you can then view the results in any pdb visualization application of your choice. One common option is to use the *PyMOL* software package [Schrödinger, LLC, 2010]. While it is not the purpose of this vignette to teach the reader *PyMOL* syntax, we present the following simplistic example and the resulting figure for reference. It will color the first cluster outputted by the **iPAC** method, residues with serial numbers 2583-2846 in blue. The chain and resSeq information provided in Example 4 is used as below.

Code Example 5: PyMOL sample code

```

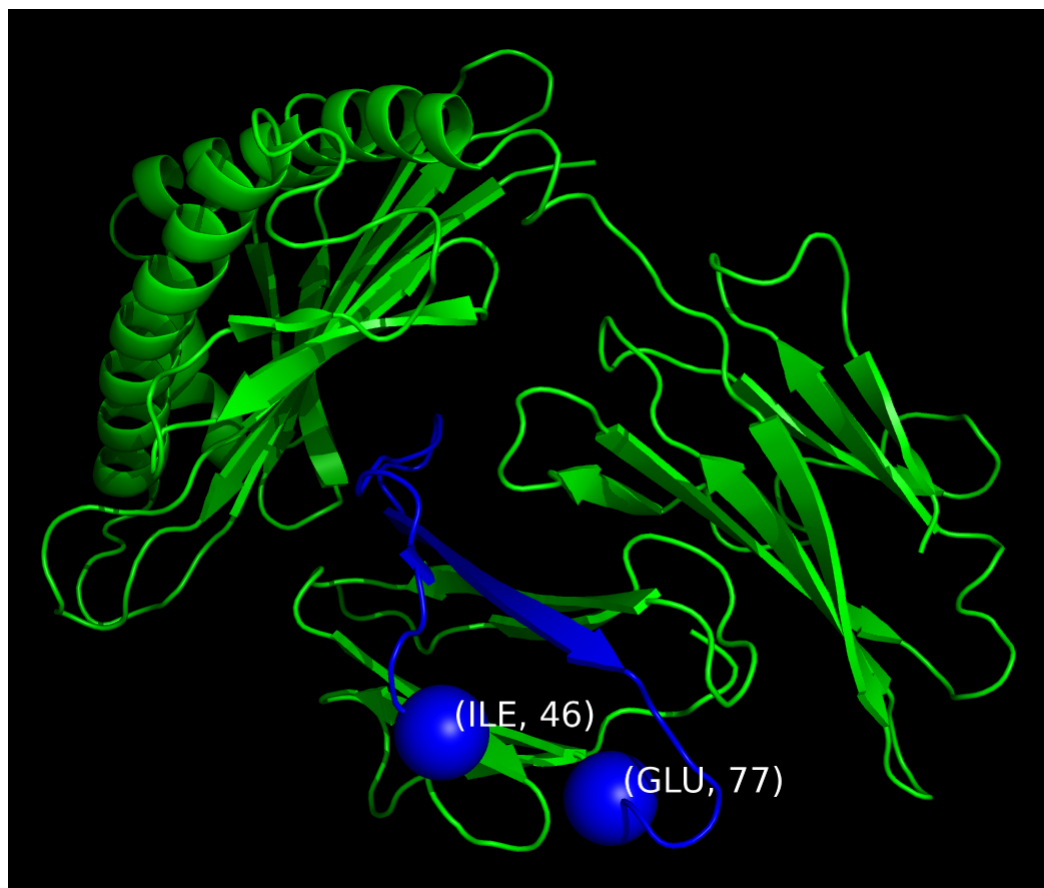
-----
hide all
show cartoon,
show spheres, ///b/46/ca
show spheres, ///b/77/ca
color blue, ///b/46-77

```

```

label c. B and n. CA and i. 46, "(%s, %s)" % (resn, resi)
label c. B and n. CA and i. 77, "(%s, %s)" % (resn, resi)
set label_position, (3,2,10)

```



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