# Moran I Coefficient Fluctuation Analysis for FYN Variations in Adherens Junction

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#### 1 Abstract

Signaling pathways for FYN a proto-oncogene at 6q21 is a Src family tyrosine kinase with orthology "tyrosine-protein kinase Fyn [EC:2.7.10.2]" with pathway (a) Sphingolipid signaling pathway (b) Phospholipase D signaling pathway (c) Axon guidance (d) Osteoclast differentiation (e) Focal adhesion (f) Adherens junction" (g) Platelet activation cell mediated cytotoxicity (h) T cell receptor signaling pathway (i) Fc epsilon RI signaling pathway (j) Cholinergic synapse (k) Prion disease (l) Pathogenic Escherichia coli infection (l) Viral myocarditis with networks of (a) Cytoskeletal regulation (viruses and bacteria), (b) Pl3K signaling and Prion disease with network elements of Escherichia EspJ/Tir to Actin signaling pathway PRNP-Pl3K-NOX2 signaling pathway and Scrapie conformation PrPSc to PRNP-Pl3K-NOX2 signaling pathway.

Variations in the structure of FYN with 30 different structures is examined here with Hasegawa, Kishino and Yano (HKY85) model and properties examined with the Protein Stability Index Binding Potential ALiphatic f.1 CpH5 CpH7 and CpH9 and ten Kidera factors. The first four factors are essentially pure physical properties; the remaining six factors are superpositions of several physical properties as well as Moran I Coefficient Fluctuation Analysis.

Keywords: tyrosine-protein kinase proto-oncogene Moran I Coefficient 6IPY 4U1P 3UA7 6IPZ 4EIK 3H0H 3UA6 1SHF 4U17 4ZNX 3H0I 1FYN 1M27 1EFN 4D8D 1G83 3H0F 2DQ7 1AVZ 5ZAU 2MRJ 2MRK 2MQI 1ZBJ 1A0N 1AZG 1AOT 1AOU 1NYF 1NYG

## 2 Introduction

FYN Variations are involved in Environmental Information Processing such as Signal transduction with Phospholipase D signaling pathway and Sphingolipid signaling pathway with Cellular Processes Cellular community - eukaryotes as Focal adhesion and Adherens junction. For Organismal Systems and the Immune system (a) Platelet activation (b) cell mediated cytotoxicity (c) T cell receptor signaling pathway (d) Fc epsilon RI signaling pathway with 09156 Nervous system Cholinergic synapse an Development and regeneration with Axon guidance and Osteoclast differentiation. [401]

Human Diseases such as Infectious disease with bacterial like Pathogenic Escherichia coli infection and Neurodegenerative disease with Prion disease, Cardiovascular disease with Viral myocarditis are involved with FYN. Metabolism Protein kinases, genetic information processing Membrane trafficking. Enzymes with 2. Transferases 2.7 Transferring phosphoruscontaining groups 2.7.10 Protein-tyrosine kinases 2.7.10.2 non-specific protein-tyrosine kinases are also connected to FYN as well as Protein kinases such as Non-receptor tyrosine kinases SRC family. Membrane trafficking with Exocytosis and Calcium ion-dependent exocytosis Kinases and associated proteins and Endocytosis with Lipid raft mediated endocytosis Flotillin-dependent endocytosis [401]

Figure 1 has the Cell-cell adherens junctions (AJs) is the most common type of intercellular adhesions and important for maintaining tissue architecture and cell polarity and can limit cell movement and proliferation. Changes in the phosphorylation state of beta-catenin affect (a) cell-cell adhesion, (b) cell migration and the (c) level of signaling beta-catenin. Wnt signaling acts as a positive regulator of beta-catenin by inhibiting beta-catenin degradation and stabilizes beta-catenin and causes its accumulation.[401]

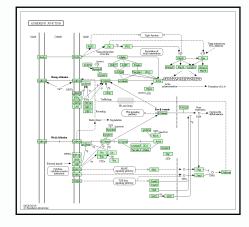


Figure 1: The KEGG graph hsa04520 has the Cell-cell adherens junctions (AJs) is the intercellular adhesions. [401]

## 3 Mathematical Model

Let  $E_1$ ,  $E_2$ ,  $E_3$ ,  $E_4$  are the states, then the transition matrix [1]

$$P(t) = (P_{ij}(t)) \tag{1}$$

where each individual entry,  $P_{ij}(t)$  refers to the probability that state  $E_i$  will change to state  $E_j$  in time t. The corresponding transition matrices for A,G,C,T is

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$$P(t) = \begin{pmatrix} p_{AA}(t) & p_{AG}(t) & p_{AC}(t) & p_{AT}(t) \\ p_{GA}(t) & p_{GG}(t) & p_{GC}(t) & p_{GT}(t) \\ p_{CA}(t) & p_{CG}(t) & p_{CC}(t) & p_{CT}(t) \\ p_{TA}(t) & p_{TG}(t) & p_{TC}(t) & p_{TT}(t) \end{pmatrix} \tag{2}$$

where the top-left and bottom-right 2x2 blocks correspond to "transition probabilities" and the top-right and bottom-left 2x2 blocks corresponds to "transversion probabilities". [1000] Table 1 has the maximum likelihood fit of sequence 1 and estimate of the eigenvalues from the markov chain transition matrix.

	1	2
1	1+0i	0.612672547758859+0i
2	0.548598035811199+0i	-0.253217867138598+0.43157696760881i
3	-0.253217867138598-0.43157696760881i	0.160762332831767+0.457336167791463i
4	0.160762332831767-0.457336167791463i	0.383223041837455+0i
5	-0.330813409838521+0i	0.21505677396784+0.243678398698828i
6	0.21505677396784-0.243678398698828i	-0.299147890266987+0.125575933714316i
7	-0.299147890266987-0.125575933714316i	-0.131874050947548+0.195229080875112i
8	-0.131874050947548-0.195229080875112i	0.091335297535122+0.201048572422311i
9	0.091335297535122-0.201048572422311i	-0.0652236932464657+0i
10	-1.5288188771906e-16+0i	1+0i

#### 4 Results

Table 2 has protein, Stability Index, Binding Potential, ALiphatic, f.1, CpH5, CpH7 andCpH9. [1001]

	Protein	Stability Index	Binding Poten- tial	ALiphatic	f.1	CpH5	CpH7	СрН9
1	6IPY	37.4	1.43	81.5	1	-4.90	-8.29	-10.032
2	4U1P	31.1	2.23	71.5	1	8.30	2.33	-2.676
3	3UA7	41.0	1.66	69.2	1	-17.00	-24.03	-27.008
4	6IPZ	34.9	1.51	68.1	1	-5.95	-7.78	-9.038
5	4EIK	55.0	1.62	80.6	1	-2.03	-3.78	-5.039
6	3H0H	34.9	1.33	74.3	1	-5.95	-7.78	-9.038
7	3UA6	39.0	1.73	70.9	1	-9.02	-12.53	-14.361
8	1SHF	37.4	1.76	72.7	1	-9.95	-13.53	-15,361
9	4U17	28.3	2.34	69.4	1	30.97	17.84	5.781
10	4ZNX	53.2	1.91	68.6	1	-8.22	-15.03	-18,009
11	3H0I	37.4	1.43	79.3	1	-11.95	-15.53	-17.360
12	1FYN	50.5	1.38	65.0	1	-4.95	-6.78	-8.112
13	1M27	28.4	1.38	86.9	1	2.92	-2.12	-6.426
14	1EFN	41.4	1.57	79.3	1	7.62	-6.70	-14.003
15	4D8D	38.2	1.43	78.0	1	5.62	-8.70	-16.033
16	1G83	34.2	2.22	72.1	1	4.32	-6.07	-10.449
17	3H0F	34.8	1.47	72.7	1	-5.95	-7.78	-9.038
18	2DQ7	40.4	1.45	89.7	1	5.37	-1.16	-6.852
19	1AVZ	39.8	1.51	79.1	1	9.68	-2.94	-9.679
20	5ZAU	33.3	1.74	71.7	1	3.75	1.23	-0.462
21	2MRJ	29.2	2.33	71.2	1	10.92	6.85	2.449
22	2MRK	31.3	2.35	68.3	1	9.14	4.85	0.375
23	2MQI	29.2	2.33	71.2	1	10.92	6.85	2.449
24	1ZBJ	37.4	1.76	72.7	1	-4.95	-6.78	-8.038
25	1A0N	49.5	1.58	70.4	1	-2.03	-3.78	-5.054
26	1AZG	49.5	1.58	70.4	1	-2.03	-3.78	-5.054
27	1AOT	36.9	2.34	75.0	1	6.48	1.95	-0.419
28	1AOU	36.9	2.34	75.0	1	6.48	1.95	-0.419
29	1NYF	40.4	1.64	74.0	1	-4.03	-5.78	-7.038
30	1NYG	40.4	1.64	74.0	1	-4.03	-5.78	-7.038

Table 1: A protein whose instability index is smaller than 40 is predicted as stable, a value above 40 predicts that the protein may be unstable. A protein have high binding potential if the index value is higher than 2.48. The relative volume occupied by aliphatic side chains (Alanine, Valine, Isoleucine, and Leucine) is a positive factor for the increase of thermostability of globular proteins. net charge of a protein sequence based on the Henderson-Hasselbalch equation for pH levels of 5 7 and 9. f.1 is the crosscovariance based on one of the sequence S at lag 1 is given by f.1 with property1 = Hydrophobicity based on the KyteDoolittle Scale and property2 = Hydrophobicity based on the Eisenberg scale.

# 4.1. Moran I Coefficient Fluctuation Analysis

The matrix weight is used as "neighbourhood" weights, and Moran's I coefficient is computed using the formula: [1007][1008]

$$I = n/S0 * (\sum_{i=1..n} \sum_{j=1..n} w_{ij} (y_i - y_m))(y_j - y_m) / (\sum_{i=1..n} (y_i - y_m)^2)$$
(3)

with yi = observations, wij = distance weight n = number of observations  $S0 = \sum_{i=1..n} \sum_{j=1..n} w_i j$ . The null hypothesis of no phylogenetic correlation is tested assuming normality of I under this null hypothesis. If the observed value of I is significantly greater than the expected value, then the values of x are positively autocorrelated, whereas if lobserved < lexpected, this will indicate negative autocorrelation. Here observed the computed Moran's I. expected the expected value of I under the null hypothesis. sd the standard deviation of I under the null hypothesis. p.value the P-value of the test of the null hypothesis against the alternative hypothesis specified in alternative is shown in Table 3. [1]

	FYN	Observed	Expected	SD	P.Value
1	6IPY	-0.00158	-0.0152	0.0104	0.192
2	4U1P	0.0333	-0.0087	0.00926	5.6e-06
3	3UA7	0.0366	-0.00383	0.00411	0
4	6IPZ	0.0235	-0.0161	0.0131	0.00256
5	4EIK	0.0722	-0.0147	0.0191	5.22e-06
6	3H0H	0.0175	-0.0161	0.00739	5.24e-06
7	3UA6	0.0163	-0.00862	0.0085	0.00342
8	1SHF	0.0133	-0.00855	0.0085	0.01
9	4U17	0.0405	-0.00326	0.00375	0
10	4ZNX	0.0281	-0.00377	0.00534	2.29e-09
11	3H0I	0.0187	-0.00855	0.00948	0.00411
12	1FYN	0.0217	-0.0141	0.0145	0.0135
13	1M27	0.0186	-0.00571	0.00459	1.23e-07
14	1EFN	0.0301	-0.00312	0.00321	0
15	4D8D	0.0273	-0.00308	0.00322	0
16	1G83	0.0721	-0.00312	0.00549	0
17	3H0F	0.0302	-0.0172	0.0147	0.00123
18	2DQ7	0.021	-0.00382	0.00307	6.66e-16
19	1AVZ	0.0327	-0.00386	0.00424	0
20	5ZAU	0.0243	-0.00833	0.00771	2.31e-05
21	2MRJ	0.0376	-0.0101	0.011	1.54e-05
22	2MRK	0.0253	-0.00917	0.00842	4.27e-05
23	2MQI	0.0496	-0.0101	0.00938	1.94e-10
24	1ZBJ	0.0218	-0.0172	0.0266	0.142
25	1A0N	0.0418	-0.0141	0.0128	1.25e-05
26	1AZG	0.0651	-0.0141	0.0127	4.14e-10
27	1AOT	0.0281	-0.00862	0.00478	1.62e-14
28	1AOU	0.0232	-0.00862	0.00484	5.08e-11
29	1NYF	0.0266	-0.0175	0.0178	0.0129
30	1NYG	0.0142	-0.0175	0.0294	0.28

Various fixed width running windows sizes can be used for summaries such as: (1) runsum, (2) runmean, (3) runmed, (4) runwtsum, (5) runq functions calculate the sum, mean, median, weighted sum, and order statistic. Figure 2 has the moving window first moment with 4ZNX, 5ZAU, and 1NYG. In the figure, the differences among this group is demonstrated with variation in the length 3 8 15 24 and 50 for the window. [1000]

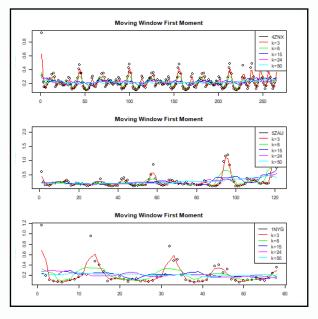


Figure 2: Moving Window First Moment for (a) 4ZNX (b) 5ZAU (c) 1NYG

Figure 3 has the Pearson Types for FYN Variations.[1000]

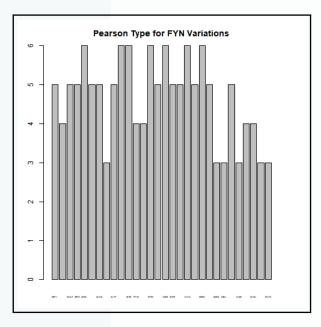


Figure 3: Pearson Types for FYN Variations

Figure 4 has the Kurtosis Values for the FYN Variations.[1000]

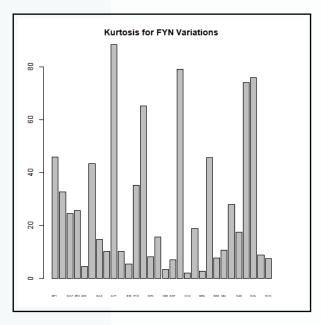


Figure 4: Kurtosis Values for the FYN Variations

### 5 Conclusions

Variations in the structure of FYN with 30 different structures is examined here with HKY85 model by Hasegawa, Kishino and Yano 1985 and properties examined with the Protein Stability Index Binding Potential ALiphatic f.1 CpH5 CpH7 and CpH9 and ten Kidera factors. The first four

factors are essentially pure physical properties; the remaining six factors are superpositions of several physical properties as well as Moran I Coefficient Fluctuation Analysis. In the moving window first moment with 4ZNX, 5ZAU, and 1NYG the variation in the length 3 8 15 24 and 50 for the window.

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