

# Prediction of Interactions between HIV-1 and Human Proteins by Information Integration

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# Human Immunodeficiency Virus-1 (HIV-1)

### Causative agent of AIDS

- Destructs the immune system
- Leads to opportunistic infections and malignancies
- Current antiviral therapy prolonged the patients' survival rates
  - Not accessible to everyone
  - Cannot eradicate HIV from the body
  - Drug resistance problems

### No vaccine



# HIV-1 Genome and Life Cycle

#### Genes Proteins





Peterlin and Trono Nature Rev. Immu. (2003) 3: 97-107

# Aim

# Predict novel direct physical interactions between HIV-1 and human proteins



# Prediction of Host Pathogen Interactions

Dyer *et al. Bioinformatics* (2007) 23(13): i159-66

- Human Plasmodium falciparum
- Co-occurrence of domain sequence signatures

Davis et al., Protein Sci (2007) 16(12): 2585-96

- Inter-PPI of human with 10 pathogens (does not include HIV)
- Comparative modeling

### General Cell (2008) 135(1): 49-60

• Functional siRNA knockout screen filtered by multiple evidences

No work to date to predict global interactome of direct physical interactions between HIV-1 and human proteins



# **Our Approach**

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X

# Supervised Learning

HIV-1 human protein pair is described with a feature vector and a class label :

$$(\bar{x}_i, y) \quad y \in \{ \text{'Interact', 'Not Interact'} \}$$

Each feature summarizes a biological information

Given data learn *a function* that would *map feature space into one of the two classes:* 

$$f: X \to Y$$

### **Training Data**



Qi et al. Proteins. (2006) 63: 490-500

Breiman Machine Learning (2001) 5-32



# Create bootstrap samples from the training data





### Construct a decision tree Use Gini Gain for splitting the nodes



# At each node in choosing the split feature choose only among *m*<*M* features





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### **Interaction** Data



# HIV-1 Human Protein Interactions

# NIAID database of human HIV-1 protein interactions curated from literature

S ncbi	HIV-1, Human Provident	r <b>otein Interaction Database</b> e of Allergy & Infectious Diseases
PubMed All Dat	abases BLAST OMIM	Map Viewer Taxonomy Structure
Search All Database	es 🗡 for	GoClear
HIV Interaction Project Gene Map Viewer RefSeq Retroviruses	env     gag     nef     p       View     all human genes that     Pr55(Gag) (is)     •       All     •       Download	ol rev tat vif vpr vpu
HIV Genotyping	All gag products	×
NIAID DAIDS	All 💙 inte	ractions
Protein Interaction DBs	Pr55(Gag) associates with	ATP-binding cassette, sub-family E, member 1 G
BIND DIP	<ul> <li>binds</li> </ul>	adaptor-related protein complex 2, alpha 1 subunit isoform 1
MINT The Binding DB	• http://ww	vw.ncbi.nlm.nih.gov/RefSeq/HIVInteraction

Sanders-Beer et al. NAR (2008) doi: 10.1093/nar/gkn708

# HIV-1 Human Protein Interactions

#### Keywords: "Nef binds hemopoietic cell kinase isoform p61HCK"

### Group 1: more likely direct interactions a)

acetylated by, acetylates, binds, cleaved by, cleaves, degraded by, dephosphorylates, interacts with, methylated by, myristoylated by, phosphorylated by, phosphorylates, ubiquitinated by

1063 interactions, 721 human proteins, 17 HIV-1 proteins

### Group 2: could be indirect interactions

activated by, activates, antagonized by, antagonizes, associates with, causes accumulation of, co-localizes with, competes with, cooperates with ...etc

1454 interactions, 914 human proteins, 16 HIV-1 proteins

HIV-1 protein

Human protein





www.hivppi.pitt.edu

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# Training and Testing Data

### The 'interaction' class:

## Group 1, the more likely direct interactions

1063 interactions, 721 human proteins, 17 HIV-1 proteins

### The 'non-interaction' class:

Select randomly from the pairs that are not reported in NIAID database

### Features



# **35** Features

Differential gene expression in HIV infected vs uninfected cells (4)

HIV-1 protein type (17)ELM-ligand feature (1)

- Human protein expression in HIV-1 susceptible tissues (1)
- Similarity of the two proteins in terms of (4)
  - Cellular location
  - Molecular process
  - Molecular function
  - Sequence

Human PPI interactome features (8)



# **ELM-Ligand Feature**

Functional interaction motifs obtained Eukaryotic Linear Motif (ELM) database

[RKY]XXPXXP

*motif involved in protein-protein interaction mediated by SH3 domains* 

# **Motif-Ligand Feature**

Functional interaction motifs obtained Eukaryotic Linear Motif database



$$f_{motif} = q$$
, where  $0 \le q \le 1$ 

Making Use of the Human PPI Interactome: Mimicry of Human Protein Interaction Partners'



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Making Use of the Human PPI Interactome: Mimicry of Human Protein's Interaction Partners



Similarity of HIV-1 protein to human protein's interaction partner

- Sequence
- Post translational modification
- Cellular location
- Molecular process
- Molecular function

Making Use of the Human PPI Interactome: Human Protein's Topological Properties the Human PPI network



**Degree** Number of neighbors



### **Clustering coefficient**

The extent the neighbors are connected with each other



 $k_{v}$ 



#### **Betweenness Centrality**

The fraction of shortest paths pass through the node



# **Evaluation**



### **Performance Measures**

### Precision Recall Curve

• Precision

- : TP/(TP+ FP)
- Recall (Sensitivity)
  - : TP/(TP+ FN)

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- The Mean Average Precision (MAP):
  - Mean of the average precisions where each average precision is calculated when recall increases.

### **Performance Measures**

### Precision Recall Curve

- Precision : TP/(TP+ FP)
- Recall (Sensitivity) : TP/(TP+ FN)
- The Mean Average Precision (MAP):
  - Mean of the average precisions where each average precision is calculated when recall increases.

Area Under the Receiver Operating Curve (AUC):



**ROC** curve

• Partial AUC scores : Area under the curve until reaching N false positives

# **Performance Evaluation**

### 10 repeated 3-fold cross validation



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**Gini importance:** Normalized sum of improvement in the "Gini gain" due a given feature in the forest

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are highly informative

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The network topology features are highly ranked

### **Network Features**

### Epstein–Barr virus targets high degree human proteins

Calderwood et al., PNAS (2007) 104: 7606-11

Pathogens tend to interact with host proteins with high degrees and betweenness centrality

Dyer et. al. PLoS Pathog (2008) 4, e32





How can we perform using only the top 6 features?

# When only the Top Ranked Features Used



Top 6 Gini Features:

- 1. Degree
- 2. Betweenness centrality
- 3. Neighbor process similarity
- 4. Clustering coefficient
- 5. Neighbor function similarity
- 6. Neighbor location similarity

# When only the Top Ranked Features Used



# When only the Top Ranked Features Used



# **Predicted Interactions**



### Predictions

Apply the model to all possible HIV-1, human protein pairs

				-		_
Score Cutoff	<b>Total Pairs</b>	Group 1	Group 2	Novel	Recall	Precision
≥ 0.00	3372	1040	232	2100	0.51	0.20
≥ 0.50	1942	1034	141	767	0.37	0.29
≥ 1.00	1440	1023	68	349	0.26	0.36
≥ 1.50	1085	894	34	157	0.18	0.41
≥ 2.00	622	538	15	69	0.13	0.47
≥ 2.50	279	243	8	28	0.09	0.47
	l	Jsed in model	Predicti	ons		

construction

# Increasing Novelty

**High Precision** 

**High Recall** 



## Functionally Interesting Interactions

- 304 cellular proteins detected in virion
  - Ott Rev Med Bio (2008 17: 159-75)
- 273 genes that had an effect in the Brass siRNA screen
  - Brass et al, Science (2008) 319: 921-6
- 295 genes that had an effect in the Konig siRNA screen
  - Konig *et al.* Cell (2008) 1: 49-60
- □ The interactors of the siRNA genes



Recall	Precision	in Virion	Brass <i>et al.</i> siRNA screen		Konig <i>et al</i> .	screen
			Genes	Interactors	Genes	Interactors
0.51	0.20	246	46	1064	77	422
0.37	0.29	101	13	441	21	181
0.26	0.36	48	5	212	11	99
0.18	0.41	17	2	99	7	53
0.13	0.47	8	1	49	4	28
0.09	0.47	4	0	25	2	14

### Tat interacts with Pin1

#### www.cs.cmu.edu/~HIV/hivPPI.html

HIV-1 protein name	Human patner Entrez gene id	Human partner gene symbol	Human partner official name	Random forest score
gag_matrix	5566	PRKACA	"protein kinası	4.34
tat	5970	RELA	"v-rel reticuloe	4.31
gag_matrix	801	CALM1	"calmodulin 1	4.30
env_gp160	801	CALM1	"calmodulin 1	4.22
nef	5566	PRKACA	"protein kinası	4.17
tat	6598	SMARCB1	"SWI/SNF rela	4.12
env_gp120	801	CALM1	"calmodulin 1	4.11
tat	3725	<u>JUN</u>	"jun oncogene	4.10
nef	7157	<u>TP53</u>	"tumor protein	4.10
nef	2534	<u>EYN</u>	"FYN oncoger	4.05
tat	5111	PCNA	"proliferating c	4.02
tat	5590	PRKCZ	"protein kinası	4.00
tat	2071	ERCC3	"excision repa	3.99
tat	2961	GTF2E2	"general trans	3.91
env_gp41	801	CALM1	"calmodulin 1	3.90
rev	1457	CSNK2A1	"casein kinase	3.90
env_gp160	2335	<u>EN1</u>	"fibronectin 1"	3.90
tat	5588	PRKCQ	"protein kinası	3.87
<u>n</u> ef	5578	PRKCA	"protein kinası	3.87
nef	801	CALM1	"calmodulin 1	3.86

detected in virion



*Pin1 interacts with and reduces expression of APOBEC3G.* Watashi JV (2008) 82: 9928-36

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# Conclusions

- Collected data from multiple biological information sources and encoded as features
- Developed a model to predict HIV-1, human protein interaction network
- Features containing human proteome knowledge is highly informative
- Specific protein interactions are being tested
- Predictions available

www.cs.cmu.edu/~HIV/hivPPI.html

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Jaime G. Carbonell Carnegie Mellon University

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We invite you to explore our web pages and learn more about our center and the biology of HV. After reading through these pages, if you have any questions or comments or would like to begin a scientific collaboration, please contact the PCHPI coordinator.<u>Teresa.Brosentsch</u>. We hope you find these pages heipful and look forward to hearing from you.



Judith Klein-Seetharaman

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### Precision/Recall Curve When Protein Type Features Excluded



Table S8. AUC scores computed in false positive range.

	AUC0.1	AUC0.05	AUC0.01	AUC0.001
Avg	0.6092	0.4958	0.2374	0.0527
Std	0.0183	0.0218	0.0235	0.0125



### The Interaction Data Counts

HIV protein	Number of HIV-1- Human Interactions			
_	Group 1 type	Group 2 type		
Envelope gp41	37	118		
Envelope gp120	195	336		
Envelope gp160	54	121		
Gag capsid	19	13		
Gag matrix	39	37		
Gag nucleocapsid	5	19		
Gag p6	14	0		
Gag pr55	15	32		
Nef	71	119		
Integrase	72	6		
Protease	60	18		
Reverse transcriptase	17	22		
Rev	33	29		
Tat	336	420		
Vif	54	10		
Vpr	35	134		
Vpu	7	13		
Total	1063	1454		
Number of unique human proteins involved	721	914		

# **Evaluation Design**

- 1. Randomly select the negative examples from non-interacting pairs
- 2. Repeated 3-fold cross validation



Repeated 10 times. The performance is average of 30 runs.

### Features

□ 35 features calculated for e HIV-1 , human protein pair



8 features specific to HIV-1, human protein pair



10 features specific to human protein



17 features specific to human protein