Makina Öğrenmesi ile Sinirsel Gelişim Hastalıkları için Gen Keşfi

A. Ercüment Çiçek

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Yol Haritası

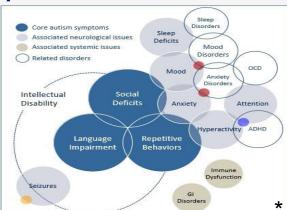
- Δ Otizm ve sinirsel gelişim hastalıkları
- Δ Gen keşfi problemi
- Δ Literatürdeki algoritmalar
- Δ ST-Steiner: Zaman-mekansal gen keşfi algoritması



Otizm

- Δ Semptomlar: Konuşma ve sosyal iletişim bozukluğu, tekrar eden hareketler
- Δ Geniş spektrum
- Δ ABD'de her 56 çocuktan 1'inde görülüyor
- Δ Genetik mimari:
- Yüksek seviyede kalıtsal
- ve heterojen





Gen Keşfi

Analysis of Rare, Exonic Variation amongst Subjects with Autism Spectrum Disorders and Population Controls

Li Liu, Aniko Sabo, Benjamin M. Neale, Uma Nagaswamy, Christine Stevens, Elaine Lim, Corneliu A. Bodea, Donna Muzny, Jeffrey G. Reid, Eric Banks, Hillary Coon, Mark DePristo, Huyen Dinh, [...], Kathryn Roeder [1] (view all]

Metrics

Published: April 11, 2013 • http://dx.doi.org/10.1371/journal.pgen.1003443

Article	Authors
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Abstract	
Author Summary	Ab
Introduction	We
Results	auti
Discussion	simi
Methods	prod
Supporting Information	stra
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Reader Comments (0)	has

PLOS GENETICS

Figures

Media Coverage (0)

Abstract

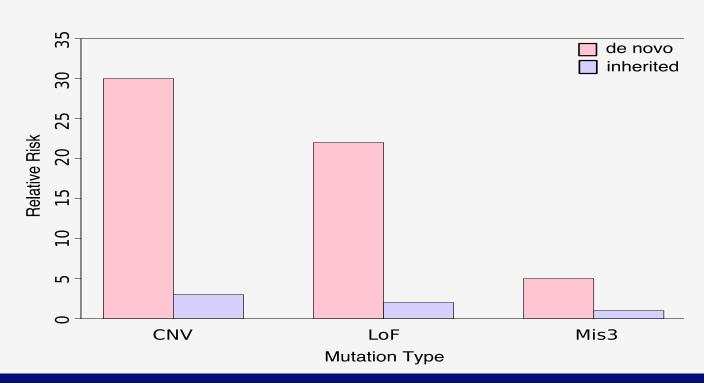
report on results from whole-exome sequencing (WES) of 1,039 subjects diagnosed with sm spectrum disorders (ASD) and 870 controls selected from the NIMH repository to be of ilar ancestry to cases. The WES data came from two centers using different methods to duce sequence and to call variants from it. Therefore, an initial goal was to ensure the ribution of rare variation was similar for data from different centers. This proved ightforward by filtering called variants by fraction of missing data, read depth, and balance Iternative to reference reads. Results were evaluated using seven samples sequenced at n centers and by results from the association study. Next we addressed how the data and/or ults from the centers should be combined. Gene-based analyses of association was an ious choice, but should statistics for association be combined across centers (metallysis) or should data be combined and then analyzed (mega-analysis)? Because of the ure of many gene-based tests, we showed by theory and simulations that mega-analysis better power than meta-analysis. Finally, before analyzing the data for association, we explored the impact of population structure on rare variant analysis in these data. Like other recent studies, we found evidence that population structure can confound case-control studies by the clustering of rare variants in ancestry space; yet, unlike some recent studies, for these data we found that principal component-based analyses were sufficient to control for ancestry and produce test statistics with appropriate distributions. After using a variety of gene-based tests and both meta- and mega-analysis, we found no new risk genes for ASD in this sample. Our results suggest that standard gene-based tests will require much larger samples of cases and controls before being effective for gene discovery, even for a disorder like ASD.

Comments

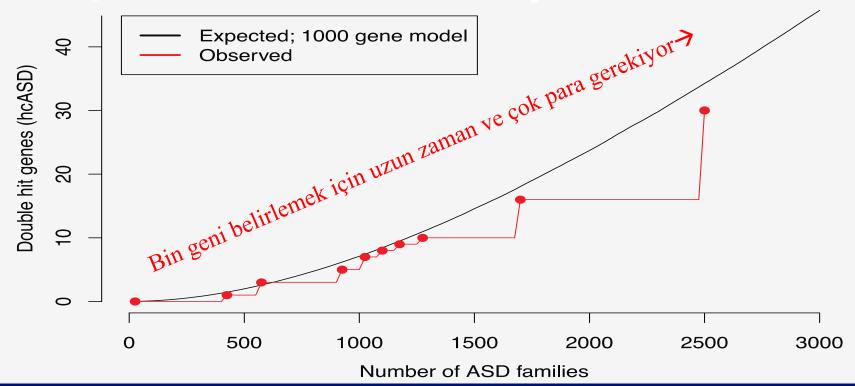
Related Content

de novo mutasyonların önemi

Mike Wigler Lab



Gen Keşfi: *de novo* LoF mutasyonlar



Gen keşfi: Son durum 2014

	(00)		(16)	
CTTNBP2	(23)	(10)	()	
BCL11A	NR3C2	POGZ		
TRIO	ASH1L	ADNP	CHD2	FOXP1
MLL3	ETFB	CHD8	WDFY3	GIGYF1
APH1A	SYNGAP1	TBR1	TNRC6B	KMT2E
ASXL3	CACNA2D3	KATNAL2	KDM6B	MED13L
MIB1	GABRB3	GRIN2B	DSCAM	DIP2A
VIL1	SETD5	ANK2	PHF2	WAC
RELN	SUV420H1	SCN2A	RIMS1	TCF7L2
CDC42BPB	NAA15	ARID1B	NCKAP1	KDM5B
MYT1L	PTEN	DYRK1A		
CUL3	MYO9B			
		*		





Gen keşfi: Son durum 2015

(10)(15)(16)CTTNBP2 **POGZ** NR3C2 BCL11A CHD2 **ADNP** FOXP1 ASH1L **TRIO** WDFY3 CHD8 GIGYF1 **ETFB** MLL3 TNRC6B TBR1 KMT2E SYNGAP1 APH1A KDM6B KATNAL2 MED13L CACNA2D3 ASXL3 **DSCAM GRIN2B** DIP2A **GABRB3** MIB1 PHF2 ANK2 WAC SETD5 VIL1 RIMS1 SCN2A TCF7L2 SUV420H1 RELN NCKAP1 ARID1B KDM5B NAA15 CDC42BPB DYRK1A PTEN MYT1L MYO9B CUL3

ASC + SSC + küçük CNV

RANBP17 (20)ZC3H4 **SPAST** SARM1 ILF2

SLC6A1 **ZNF559 MFRP** P2RX5 NINL

INTS6 USP45 CAPN12 **OR52M1**

AKAP9

TRIP12 IRF2BPL **ACHE**

PTK7 **ERBB2IP**

NRXN1 (7) SHANK2 SHANK3 **EP400**

KAT2B STARD3NL **NLGN3**

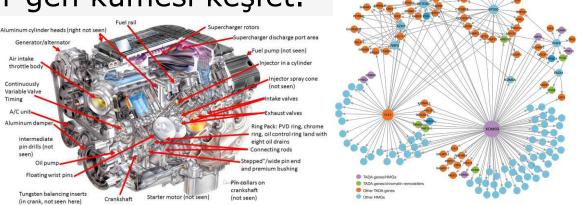
Gen Keşfi için algoritmalar

Kabullenim: Risk genleri biyolojik ağlarda fonksiyonel bir küme oluşturur.

Guilt-by-association prensibi:

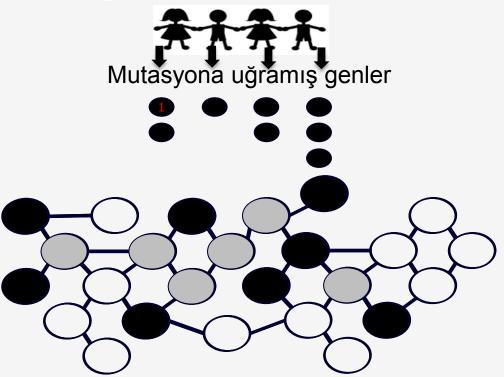
1. Bilinen ASD genlerini başlangıç noktası olarak kullan.

2. Sıkı ilişki içinde bir gen kümesi keşfet.





Gen Keşfi için algoritmalar



Gen Keşfi için algoritmalar

- **NETBAG¹**, NETwork-Based Analysis of Genetic Associations
- **DAWN**², Detecting Association With Networks
- MAGI³, Merging Affected Genes into Integrated networks
- **ST-Steiner**⁴, Spatio-Temporal Gene Discovery for ASD

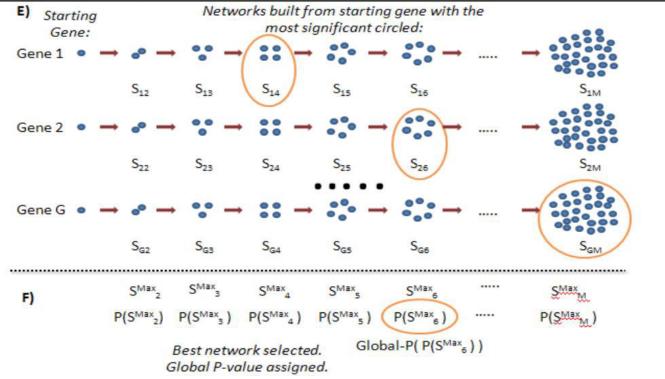
¹ Gilman et al. 2011, "Rare de novo variants associated with autism implicate a large functional network of genes involved in formation and function of synapses". *Neuron*.

² Liu et al. 2014, "DAWN: a framework to identify autism genes and subnetworks using gene expression and genetics". *Molecular Autism*.

³ Hormozdiari et al. 2015, "The discovery of integrated gene networks for autism and related disorders". *Genome Research*.

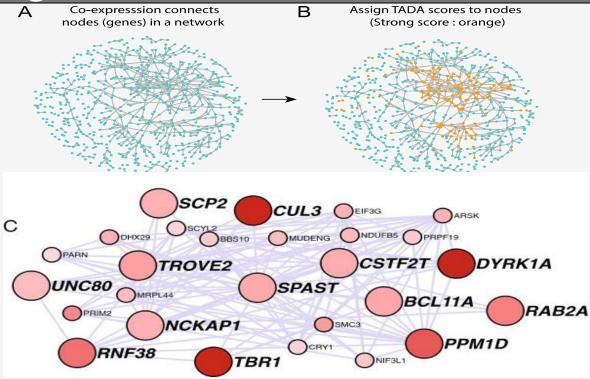
⁴ Norman and Cicek, 2018, Spatio-Temporal Gene Discovery for Autism Spectrum Disorder". *bioRxiv* 2018.

NETBAG Algoritması

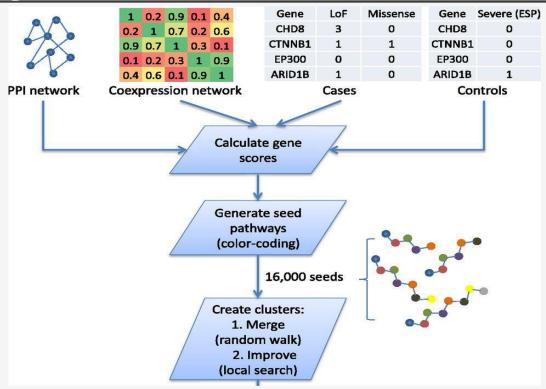




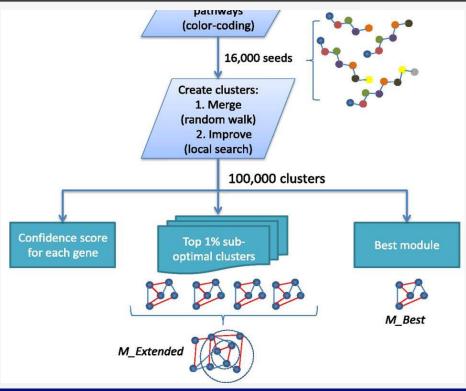
DAWN Algoritması



MAGI Algoritması



MAGI Algoritması



Eksiklikler

Gen etkileşimleri zamanla değişir

Farklı beyin gelişim pencereleri:

- Farklı topolojilere sahiptir,
- ASD gen kümelenmesi de zaman ve mekana göre değişir.

Significance of observed pASD enrichment by brain region and period of development - hypergeometric test

1.7

8.0 2.9

Periods of development (temporal)

Willsey et. al., 2013, Cell 155.

Erken dönemlerde, biyolojik patikalardaki problemler ileriki dönemlere ardışık olarak etki eder.

- Erken donem patika hasarı ileri dönemleri etkiler ve 36 kat
 ASD risk artısına neden olur*
- Yetişkinlikteki problemler ise risk artışına neden olmaz*.



Eksiklikler

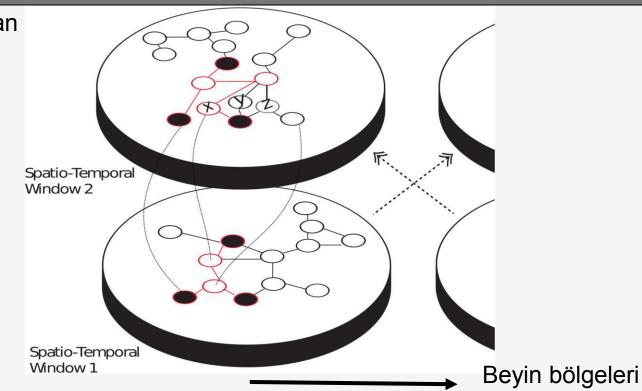
Bu nedenle statik biyolojik ağlar kullanan metotlar: **Dinamik beyin gelişimini modelleyemez** ve tahmin güçleri limitlidir.

Hipotezimiz:

"Gen kümelenmesi statik değil dinamiktir."

ST-Steiner Algoritması

- Zaman O Gene
- O Steiner Tree Gene
- ASD Risk
 Gene
- Coexpression
- Steiner Tree edge
- Same Gene in consequtive windows
 - Cross
- --»time/brain region interaction



Methodlar

Statik biyolojik ağların yarattığı sorunları aşmayı hedefliyoruz:

- Prize-collecting Steiner tree (PCST) problemi kullanacağız.
- Zaman ve mekan bilgisini de algoritmaya vereceğiz.

Orijinal Steiner tree (ağaç) problemi: Verilen bir ağda, ayrıcalıklı, tohum noktaları bağlayan minimum bir ağaç bulur.



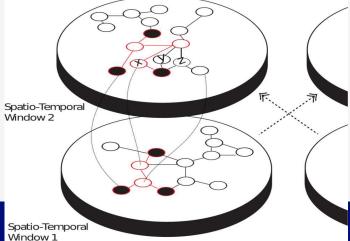
Prize-Collecting Steiner Tree (PCST)

Prize-collecting Steiner tree problemi:

Öyle bir ağaç bul ki:

Seçilen nodların ödülleri maksimum.

Seçilen bağlantıların cezaları minimum olsun.



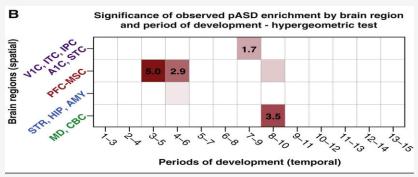
Prize-Collecting Steiner Tree (PCST)

 $G(V, E)_{\mathbf{r}}$ a nodları ve bağlantıları ağırlıklı bir ağ olsun Nod seti V, bağlantı seti E, bağlantı cezası fonksiyonu $c(e) \ge 0$, nod ödülü fonksiyonu $p(v) \ge 0$

Aranan ağaç $T(V_T, E_T)$ bu fonksiyonu minimize eder:

$$o_T(T) = \sum_{e \in E_T} c(e) + \beta \sum_{v \notin V_T} p(v) \quad \beta \ge 0$$

Problem Formulasyonu



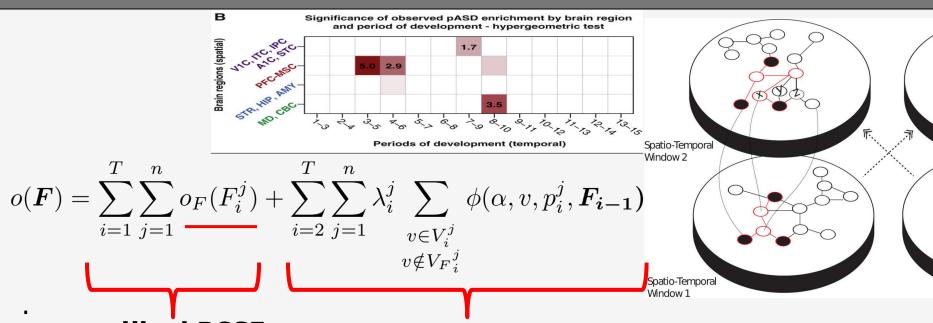
Willsey et. al., 2013, Cell 155.

- Zaman-mekânsal sistem
- i. pencere,

- $G=(~G_1,G_2,\cdots,G_T)$, T zaman penceresi $G_i=\{~G_i^1,G_i^2\cdots,G_i^n\}$ n tane beyin bölgesi içerir.
- Buradan bir alt sistem bulmak istiyoruz.

$$m{F} = (m{F_1}, m{F_2}, \cdots, m{F_T})$$
 $m{F_i} = \{ m{F}_i^1, F_i^2 \cdots F_i^n \}$ $m{F}_i^j(V_F{}_i^j, E_F{}_i^j)$ alt kümesidir $m{G}_i^j(V_i^j, E_i^j)$

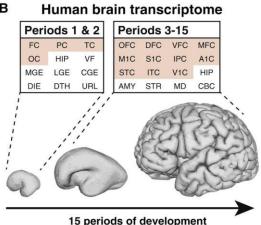
ST-Steiner



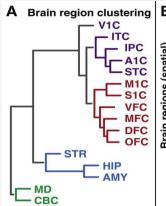
orijinal PCSF çözümleri

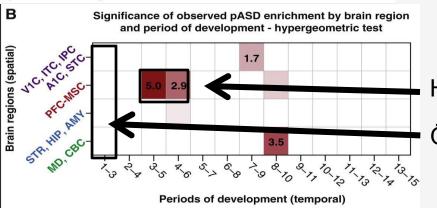
farklılıkları cezalandıran fonksiyon

Veri setleri



Period	Description	Age	
1	Embryonic	4-8 PCW	
2	Early fetal	8-10 PCW	
3	Early fetal	10-13 PCW	
4	Early mid-fetal	13-16 PCW	
5	Early mid-fetal	16-19 PCW	
6	Late mid-fetal	19-24 PCW	
7	Late fetal	24-38 PCW	
8	Neonatal & early infancy	0-6 M	
9	Late infancy	6-12 M	
10	Early childhood	1-6 Y	
11	Middle and late childhood	6-12 Y	
12	Adolescence	12-20 Y	
13	Young adulthood	20-40 Y	
14	Middle adulthood	40-60 Y	
15	Late adulthood	60 Y+	





Hedef Önceki ağlar

Veri setleri

ASC WES¹ (Öğrenme)

- 16,098 DNA örneği
- 3,871 ASD örneği (trio)
- 9,937 kontrol örneği

SSC WES² (Test)

- 1,643 ek DNA örneği (trio)
- 251 vurulmuş gen

Bu 251 geni tahmin edebilir miyiz?



Sonuçlar 1

gene set name	p-value	intersection $/$ # genes predicted
ST-St. Every $(1-3)$ +PFC $(3-5)$	6.958e-12	$21 \; / \; 234$
ST-St. Every $(1-3)$ +PFC $(4-6)$	1.827e-09	$19\;/\;256$
NETBAG	2.476e-06	9 / 87
$\overline{\text{DAWN PFC}(3-5)}$	3.842e-08	17 / 246
DAWN $PFC(4-6)$	9.017e-10	$18 \ / \ 218$
MAGI Best1	3.679e-05	6 / 47
MAGI Ext1	7.865e-05	8 / 104
MAGI Best2	2.628e-02	2 / 19
MAGI Ext2	4.396e-03	5 / 80

Alakalı Otizm Gen Listeleri

- SFARI Category 1 (24 gen)
- SFARI Category 2 (59 gen)
- FMRP Hedefleri (842 gen)
- RBFOX peak (1048 gen)
- RBFOX splice (587 gen)
- Histone Modifiers (152 gen)
- Sinaptik Genler (878 gen)



Sonuçlar 2

Predicted Gene Set S	SFARI Ca	ategory 1 SFARI Category 2		FMRP Targets		RBFOX - peak		RBFOX - splice		
	p-value	$ \cap / S $	p-value	$ \cap / S $	p-value	$ \cap / S $	p-value	$ \cap / S $	p-value	$ \cap / S $
ST-St. Every(1-3)+PFC(3-5)	1.397e-25	16 / 234	2.906e-22	19 / 234	3.198e-22	52 / 234	2.196e-09	38 / 234	1.120e-11	31 / 234
ST-St. Every(1-3)+PFC(4-6)	8.334e-23	15 / 256	5.890e-20	18 / 256	8.684e-27	60 / 256	1.735e-11	44 / 256	3.649e-08	27 / 256
NETBAG	5.062e-18	10 / 87	1.422e-13	10 / 87	1.750e-10	21 / 87	2.870e-04	14 / 87	8.199e-05	11 / 87
DAWN PFC(3-5)	5.479e-19	13 / 246	4.242e-25	21 / 246	3.985e-18	48 / 246	2.740e-09	39 / 246	3.774e-09	28 / 246
DAWN PFC(4-6)	7.067e-24	15 / 218	1.252e-19	17 / 218	7.762e-19	46 / 218	2.735e-10	38 / 218	2.231e-08	25 / 218
MAGI Best1	3.554e-07	4 / 47	9.615e-03	2 / 47	1.031e-08	14 / 47	3.943e-03	8 / 47	5.946e-04	7 / 47
MAGI Ext1	1.866e-07	5 / 104	3.220e-04	4 / 104	3.600e-06	17 / 104	1.500e-02	12 / 104	2.137e-05	13 / 104
MAGI Best2	2.652e-04	2 / 19	1.000e+00	0 / 19	2.321e-12	12 / 19	3.127e-03	5 / 19	2.512e-03	4 / 19
MAGI Ext2	3.064e-06	4 / 80	2.049e-03	3 / 80	1.165e-08	18 / 80	1.442e-03	12 / 80	3.735e-05	11 / 80

Sonuçlar 3

Predicted Gene Set S	Histone Modifiers		Synaptic	Genes
	p-value	$ \cap / S $	p-value	$ \cap / S $
ST-St. Every(1-3)+PFC(3-5)	4.709e-07	12 / 234	1.195e-04	25 / 234
ST-St. Every(1-3)+PFC(4-6)	7.890e-06	11 / 256	7.622e-05	27 / 256
NETBAG	7.103e-04	5 / 87	2.409e-03	11 / 87
DAWN PFC(3-5)	1.835e-04	9 / 246	3.274e-04	25 / 246
DAWN PFC(4-6)	7.347e-05	9 / 218	1.744e-03	21 / 218
MAGI Best1	2.100e-06	6 / 47	2.813e-01	4 / 47
MAGI Ext1	2.072e-06	8 / 104	9.394e-03	11 / 104
MAGI Best2	1.000e+00	0 / 19	6.258e-08	9 / 19
MAGI Ext2	1.000e+00	0 / 80	3.862e-06	15 / 80

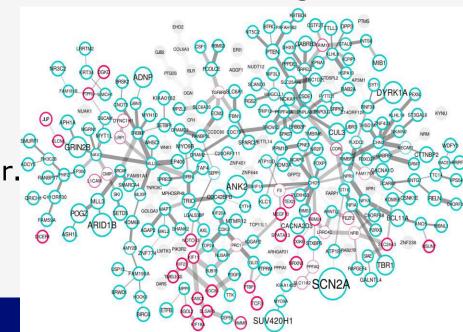


Zaman bilgisi ile elde edilen yeni genler

• 5 Kinesin ailesi (KIF) genleri: "transport cargo to dendritic spines undergoing synaptic plasticity over microtubules also play a role in organization of spindle microtubules during mitosis."

 NDC80 veSGOL2: kinetochore & microtubule attachment - cell division.

→Literatürde dile getirilmeyen genler.

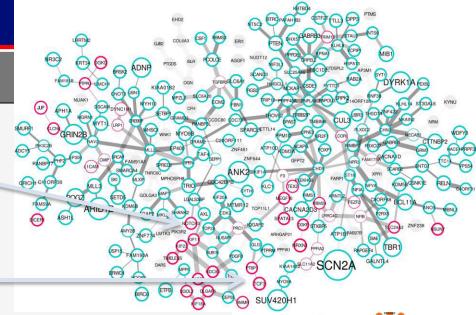


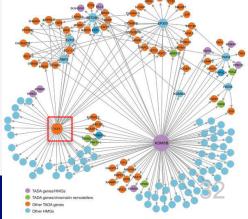


Yorumlanabilir sonuçlar

- NOTCH3
 - 💥 Nöron başkalaşımı

- TCF3
 - Embriyonik kok hücre başkalaşımını başlatır
 - Nöron öncesi hücrelerde başkalaşımı durdurur
 - De Rubeis et. al., 2014, Nature yayınında merkez gen





Çıkarımlar

Otizmin genetik mimarisinin anlaşılması önemli bir problem.

Biyolojik ağ bazlı algoritmalar hangi genlerin ilgili olduğunu tespit etmede önemli katkıda bulunmuştur.

ST-Steiner algoritması bunları bir adım öteye götürerek zaman mekânsal değişiklikleri hesaplamalarına katmıştır.

ST-Steiner daha hassas sonuçlar vermektedir:

- Bağlantı için gerekli olmayan genleri dışarıda bırakır.
- Önceki donemde seçilen genleri hesaba katarak daha güvenli tercihler yapar.
- Sonuçlar kümelenmenin statik değil dinamik olduğunu desteklemektedir.





Call for submissions opens

Submission deadline

Acceptance notification

Early registration discount ends

Conference Program

ISCB-RSG Turkey – Student Symposium

May 15, 2018

August 1, 2018

September 1, 2018

September 15, 2018

October 25-27, 2018

October 27, 2018

ISCB-RSG Turkey - Student Symposium

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Teşekkürler!

Spatio-Temporal Gene Discovery for Autism Spectrum Disorder



Utku Norman

Bilkent Yüksek Lisans mezunu - EPFL doktora öğrencisi















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