Bayesian networks for DNA-based kinship analysis Functionality and validation of GENis missing person identification module



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Abstract

GENis is a recently published open-source multi-tier information system developed to run forensic DNA databases. It relies on a Bayesian Networks framework and it is particularly well suited to efficiently perform large-size queries against databases of missing individuals. In this contribution of the missing person identification capabilities of **GENis**. To that end we introduce **fbnet**, a free-software package written in the R statistical language that implements the complete **GENis** functionality to perform kinship analysis based on DNA profiles. With the aid of **fbnet**, we could validate likelihood ratios against estimations draw with Familias and forrel (two well-recognized R packages for kinship quantification) for complex pedigrees provided by the Argentinian reference databank (Banco Nacional de Datos Geneticos, BNDG).

GENis & fbnet



GENis is a highly customizable system composed of three different modules related to: (a) person identification and analysis of forensic evidence, (b) missing person identification, and (c) disaster victim identification¹. The missing person identification (MPI) module was specifically developed to run automatic queries on family and missing person databases.

For each family, the pedigree structure and available genotypes are into a bayesian network² used to infer genotype integrated probability tables for the queried missing person (MP). The availability of these probability tables allows for a rapid calculation of likelihoods



We implemented the complete GEN is functionality to perform kinship analysis in **fbnet**³, an open-source package written in R statistical language.

	> library(fbnet)							
	>pbn <- initBN(ped)	Get dat	a from <i>pa</i>	r <i>amlink</i> pe	digree			
	> bnet <- buildBN(pbn,MP=7)	Build the bayesian network associated to					digree	
	> bnet <- buildCPTs(bnet)	Compute Conditional Probability Tables						
	> resQ <- velim.bn(bnet)	Compute MP allelic probabilities conditioned by evidence MP genotype probabilities $P(MP ev,\pi)$					lence	
	> IProbGeno <- genotypeProbTable(resQ)							
	Paternal and maternal allelic probability distributions for MP		> resLR<	LR estii	mation .R(bn1,	resQ,ge	eno)	
0.0 0.2 0.4	 population fbnet 							
0.20 0.30			CSF1P0 D1	L ₀ 1.42e-13	L ₁ 4.66e	LI -14 0.32 L ₁	R 27 LR	-

for large MP databases.





···· CSF1P0 3.12e-13 4.56e-12 14.6

Validation Exercise

To speed-up calculations **GENis** and **fbnet** implement a truncated version of the canonical stepwise model where probabilities for rare mutations, i.e. more than a given number (L) of steps away from the original allelic value (i.e. the diagonal term), are neglected.

We estimated with **fbnet** LR values for 1000 unidentified persons (UPs) simulated profiles (15 markers) considering H1: UP is MP, and 1000 UPs considering H0: UP is not MP. We used L=1, 2 and 4 models. In Panel A we show the precision of these estimations, at a given tolerance level, compared against Familia's LR values for familyFF (Panel C).

LR estimations for large databases can be done very efficiently as the MP genotype probabilities, conditioned by the available evidence and pedigree relationships, are estimated only once for each family. Computational time for **fbnet** LR estimations for the ensemble of simulated profiles are compared against forrel running times in Panel B (cpu: ryzen 5 2600x 3.6Ghz, 32 gb ram)



CSF1PO paterna

Systematic comparisons of LR values

We considered 24 familial pedigrees from BNDG and randomly generated the genotype (23 molecular markers) of available family contributors. For each pedigree we conditionally simulated 1000 MP genotypes and compared the obtained LR values against forrel estimations for both, equal and stepwise mutational models (TSMM with L=4).

	#members	#genotyped	$\operatorname{Corr}\operatorname{Log}[\operatorname{LR}]$	mean error $[\%]$	$\max \ \mathrm{error} \ [\%]$	# dev>1%	# dev > 2%	# dev > 5%
F1	12	5	0.99999269	0.17	74.50	10	2	1
$F1^*$	9	3	0.99999794	0.08	14.42	5	3	2
F2	16	9	0.99999994	0.06	2.61	5	1	0
$F2^*$	12	6	0.99999991	0.06	1.86	5	0	0
F3	9	5	0.99999988	0.09	6.42	6	4	1
$F3^*$	9	5	0.99999932	0.07	13.18	6	2	1
$F4^*$	6	3	0.99999981	0.04	5.43	2	2	1
F5	9	5	0.99999998	0.04	1.79	3	0	0
$F5^*$	9	5	0.9999999	0.05	2.33	4	1	0
F6	9	3	0.99999963	0.11	9.80	14	5	2



Conclusions

We found an excellent agreement between our LR estimations and the corresponding reference values. Moreover, the bayesian network approach to the kinship analysis problem provided an extremely efficient methodology to evaluate LRs for large MP databases. Noticeably, the functionality implemented in the open-source package **fbnet** allows us to share with the community the functionality and main design principles behind GENis MPI module

*paternal branch excluded

[1] "GENis, an open-source multi-tier forensic DNA information system", Chernomoretz et al, Forensic Science International: Reports (2020) http://doi.org/10.1016/j.fsir.2020.100132 [2] "Modeling and reasoning with bayesian networks", Darwiche A., Cambridge University Press (2009). [3] fbnet R package can be downloaded from : https://CRAN.R-project.org/package=fbnet



