

No Gene Is an Island: The Flip-Flop Phenomenon

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An increasing number of publications are replicating a previously reported disease-marker association but with the risk allele reversed from the previous report. Do such “flip-flop” associations confirm or refute the previous association findings? We hypothesized that these associations may indeed be confirmations but that multilocus effects and variation in interlocus correlations contribute to this flip-flop phenomenon. We used theoretical modeling to demonstrate that flip-flop associations can occur when the investigated variant is correlated, through interactive effects or linkage disequilibrium, with a causal variant at another locus, and we show how these findings could explain previous reports of flip-flop associations.

Associations of opposite alleles at the same biallelic locus with the same disease are confusing findings, particularly in the same ethnic group. For example, both the long and short allele at the *HTTLR* locus of the serotonin transporter gene have been found to be risk alleles for autistic disorder (MIM #209850) in different studies.^{1–6} Additionally, two independent studies report the opposite alleles at the same SNP in the glutathione-S-transferase omega-1 gene (*GSTO1* [MIM #605482]) positively associated with age at onset (AAO) in late-onset Alzheimer disease (AD [MIM #104300]).^{7,8} Another example is the association of the catechol-O-methyltransferase gene (*COMT* [MIM #116790]) with schizophrenia (SCZD [MIM #181500]).^{9–13} Theoretically, such “flip-flop” associations for a genuine causal variant would be unlikely to occur in samples ascertained similarly from a common population. Thus, flip-flop associations found in apparently similar samples are often regarded as spurious findings.

Flip-flops of risk alleles have also been reported across different ethnic groups^{14,15} and may be more easily explained by population differences. These flip-flop associations may indicate heterogeneous effects of the same variant that are due to differences in genetic background or environment. It is well known that changing the genetic background (strain) in transgenic mouse experiments can significantly change the resulting phenotype. Differences in linkage disequilibrium (LD) between populations could also lead to inconsistent patterns of association when non-causal variants are tested. Despite these possible explanations, it remains arguable whether flip-flop associations in general suggest a confirmation or spurious association.

False-positive results may account for most cases of irreproducibility of association studies.¹⁶ It has been speculated that extreme positive or opposite findings tend to attract investigators and journal editors and are hence published earlier than other results.¹⁷ Inconsistent findings, therefore, may be attributable to publication biases (i.e.,

significant results tend to be favored for publication) and reporting biases (i.e., investigators may report only positive findings). Lack of acknowledgment of these biases may result in false-positive results.

When multiple loci act in concert to cause a disease, a single-locus association may be confounded by the other loci. We hypothesized that some flip-flop associations may be attributable to failure to account for multilocus effects or correlation with other loci. Our study consists of three parts. In the part I, we used theoretical models to evaluate how multilocus effects can explain flip-flop associations. In part II, we examined how variable LD patterns within the *COMT* gene might lead to inconsistent associations of the gene with SCZD. In part III, we extended the framework of LD to the correlation between two nonsyntenic loci. We used observed flip-flop associations between the glyceraldehyde-3-phosphate dehydrogenase gene (*GAPDH* [MIM #138400]) and AD as an example.

Part I: Assessment of single-locus association in the presence of multilocus effects.—Suppose we conduct a single-locus association analysis in a sample of unrelated cases and controls. We observe genotypes at a target SNP denoted by “SNP1”; another untested SNP is denoted by “SNP2.” Assume that SNP1 has two alleles A and a and that SNP2 has two alleles B and b. We computed the correlation coefficient between alleles A and B, such that the frequencies

$$r = \frac{D}{\sqrt{P_A P_a P_B P_b}},$$

where $D = P_{AB} - P_A P_B$ is the LD coefficient, P_{AB} is the frequency of haplotype AB, and P_A , P_a , P_B , and P_b are allele frequencies for the four alleles A, a, B, and b, respectively. The constraints are such that $P_A + P_a = 1$, $P_B + P_b = 1$, and D lies between $-\min(P_A P_B, P_a P_b)$ and $\min(P_A P_a, P_B P_b)$.

Assuming that only SNP1 is analyzed in an association study, we evaluated how the unexamined SNP2 may influ-

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ence the observed direction of association between SNP1 alleles and risk of disease. To assess the observed association between SNP1 and disease risk in the presence of a joint effect of SNP1 and SNP2, we compared the frequency of allele A at SNP1 in affected individuals and unaffected individuals for four different theoretical models of multilocus effects (fig. 1). Models I, II, and III are based on the locus-locus interaction models proposed by Marchini et al.¹⁸ The values for parameters f , θ_1 , and θ_2 were used to compute probabilities of being affected with a disease (i.e., penetrance) for individuals carrying different two-locus genotypes. Baseline penetrance f is the probability of being affected if the two-locus genotype is $aabb$. θ_1 and θ_2 reflect the genetic effect of allele A and allele B, respectively. For example, under model I (multiplicative model), each copy of allele A increases the risk of disease by a factor of $1 + \theta_1$. Similarly, each copy of allele B increases the risk of disease by a factor of $1 + \theta_2$. We added a fourth model, model IV, that specifies simple additive effects for the two

loci. Figure 1 shows the penetrance values for various two-locus combinations under the four different models of multilocus effects. To compute the penetrance values, we specified $f = 0.001$, $\theta_1 = 0$ or 0.1, and let $\theta_2 = 0.1, 0.2$, or 0.4, such that either of the two causative alleles A and B, at best, exerts modest effect on susceptibility to disease. Using the penetrance values, we computed the odds ratio (OR) for allele A in affected versus unaffected individuals, given the allele frequencies for two alleles A and B in the whole population were either 0.1 or 0.5.

ORs for allele A at SNP1 in affected individuals versus unaffected individuals corresponding to various r values with fixed θ_1 and θ_2 are presented in figure 2 (only models I and IV are shown, since models II and III showed no flip-flop associations). Figure 2A and 2B shows how association with allele A at SNP1 varies with r under model I. Figure 2C and 2D shows how association with allele A at SNP1 varies with r under model IV. The direction of single-locus association flips (i.e., the OR crosses 1) under models

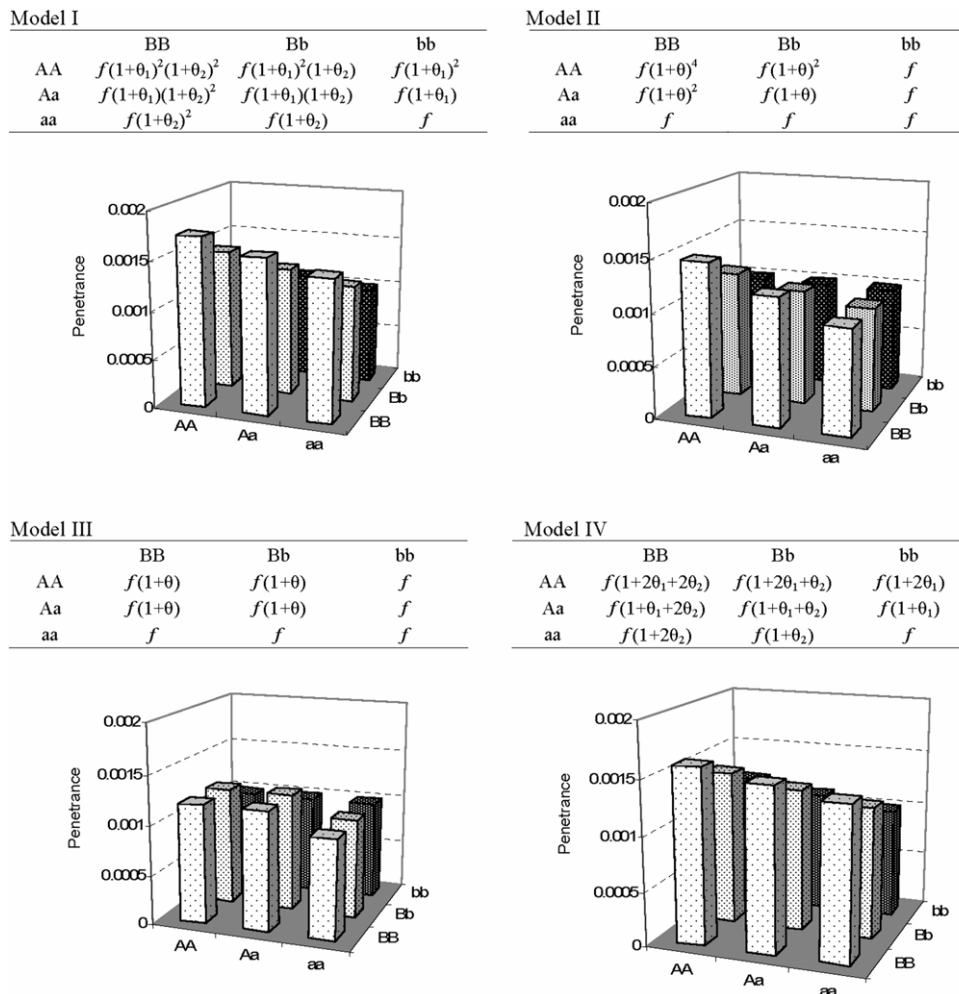


Figure 1. Penetrance values for each two-locus genotype combination under different models of multilocus effect. The penetrance values were calculated by specifying $\alpha = .001$, $\theta_1 = 0.1$, and $\theta_2 = 0.2$ for each model.

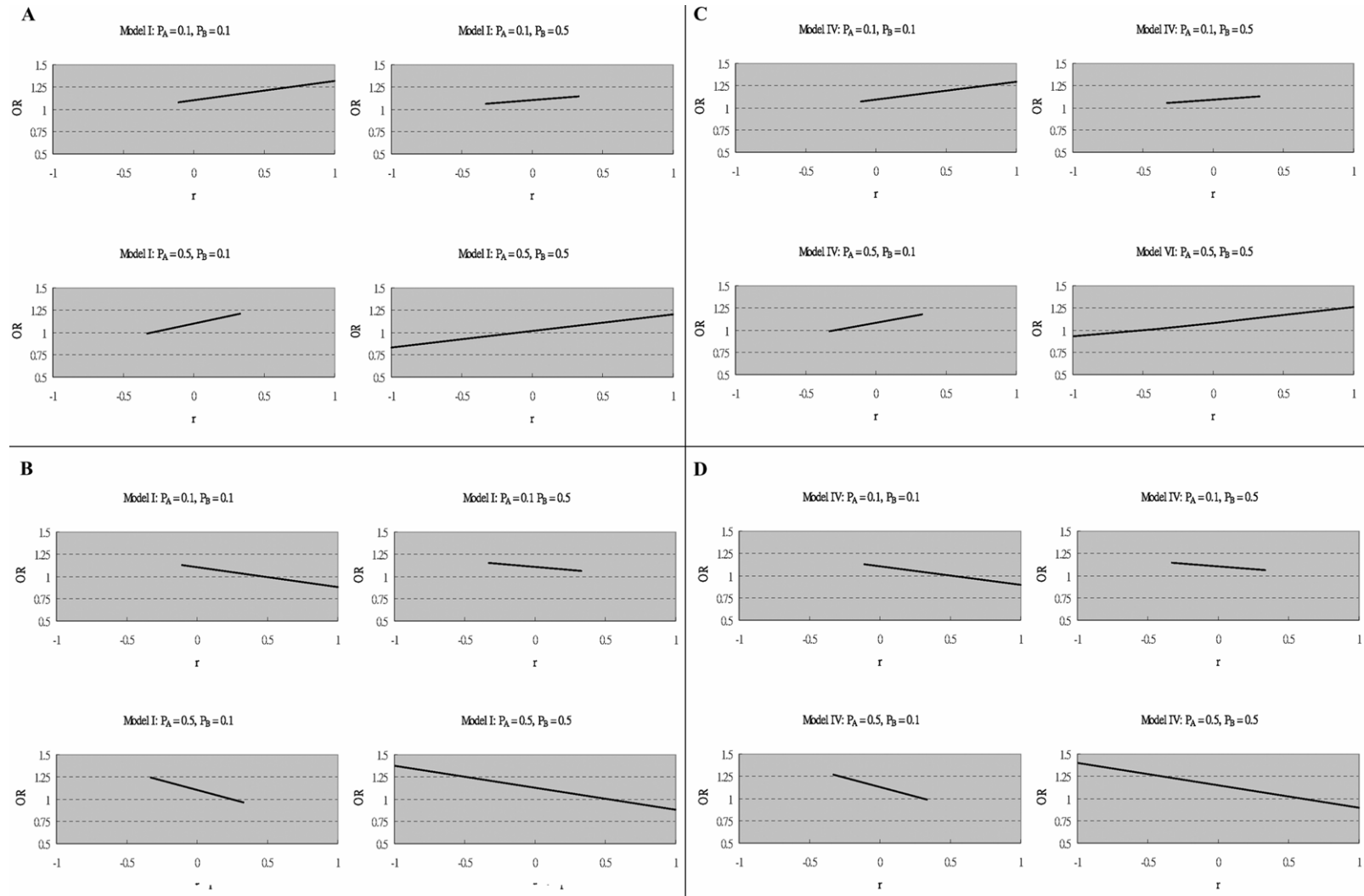


Figure 2. ORs corresponding to the effect size of allele A under models I and IV. Parameters for each situation are specified. *A*, $f = 0.001$, $\theta_1 = 0.1$, and $\theta_2 = 0.2$ under model I. *B*, $f = 0.001$, $\theta_1 = 0.1$, and $\theta_2 = -0.2$ under model I. *C*, $f = 0.001$, $\theta_1 = 0.1$, and $\theta_2 = 0.2$ under model IV. *D*, $f = 0.001$, $\theta_1 = 0.1$, and $\theta_2 = -0.2$ under model IV.

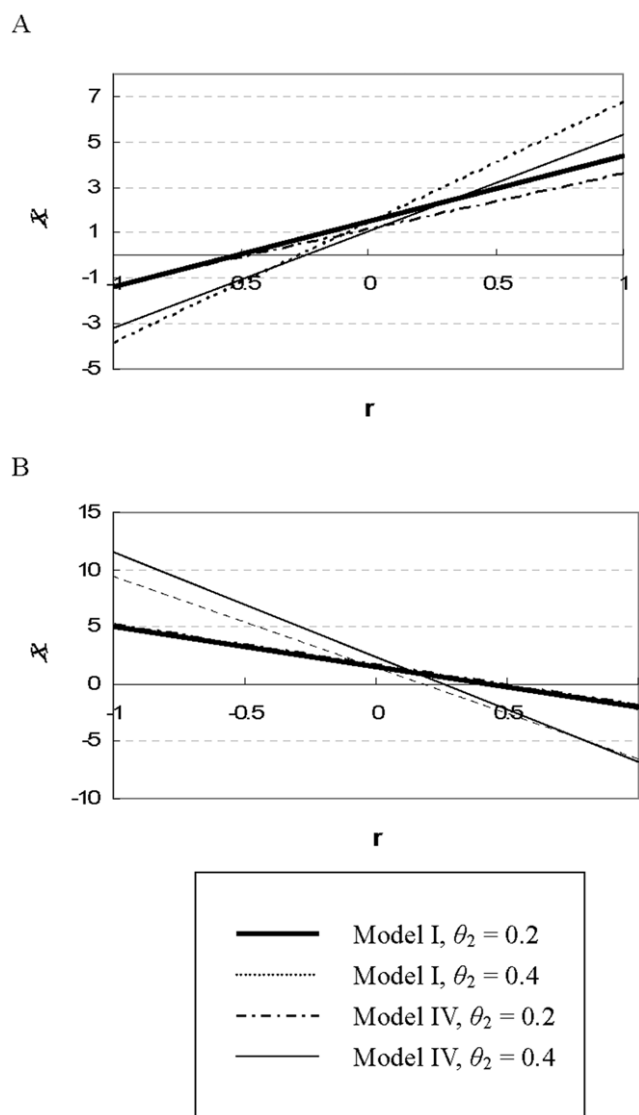


Figure 3. Directions of allelic association for the A allele in different situations. We used the statistic $\chi = (P_A - P_B) / \sqrt{P(1-P)/N}$, where $P = (P_A + P_B)/2$, to demonstrate how the direction of allelic association varies depending on θ_2 , given the same frequency of the A (and B) allele (~50%) and model of multilocus effect. θ_1 is fixed at 0.1 for all models. Panel A indicates the situation where A is a risk allele and B is also a risk allele; panel B indicates the situation where the A allele is a risk allele and the B allele is a protective allele.

I and IV as r varies. We also performed simulations in which the allele A frequency was set at 0.01 and 0.005 and saw no change in the direction of association when allele B frequency was 0.005, 0.01, 0.3, or 0.5 (data not shown). Therefore, such flip-flop associations are seen particularly when the risk allele (A) at the target locus is a relatively common allele. Additionally, the point at which the direction of association changes tends to be when the

two SNPs are in weak LD with each other (e.g., $r \approx -0.25$ in fig. 2A). This indicates that the observed direction of allelic association may be particularly susceptible to sampling variation when there is little LD, and remarkable variation in LD patterns is not required.

Figure 3 demonstrates how association of allele A with disease risk varies by r , θ_1 , and θ_2 (only models I and IV are shown). Again we see that the A allele may be positively or negatively associated with the disease, depending on the allelic correlation and the disease model. Interestingly, we may still observe a flip-flop association at SNP1 even if it has no direct effect on disease risk. This was seen in models with $\theta_1 = 0$ for which the observed SNP acts only through LD with SNP2. The results based on $\theta_1 = 0$ are similar to those based on $\theta_1 = 0.1$ (data not shown).

Part II: Examining population variation in LD across the COMT gene.—For 32 populations, we estimated, on the basis of published data,¹⁹ the correlation coefficient r between an RFLP (*Hind*III-1217) located in the P2 promoter

Table 1. Population Variation in LD between the Val158Met Polymorphism (A Allele) and the P2 Promoter (B Allele) within the COMT Gene

Population	2N ^a	Allele Frequency		r
		A	B	
Biaka	140	.51	.94	-.03
Mbuti	76	.67	.80	-.07
Yoruba	152	.86	.64	.03
Ibo	94	.66	.64	.01
Hausa	76	.57	.74	.20
Ethiopians	64	.64	.67	.03
African Americans	180	.57	.67	.19
Yemenite Jews	86	.69	.62	.66
Druze	148	.65	.62	.32
Adygei	108	.72	.66	.40
Russians	96	.52	.49	.38
Finns	70	.56	.47	.10
Danes	102	.43	.39	.11
Irish	230	.59	.50	.48
European Americans	186	.53	.38	.34
Chinese	120	.69	.78	.52
Taiwanese	98	.63	.75	.59
Hakka	80	.65	.83	.46
Japanese	102	.58	.69	.31
Ami	80	.75	.85	.41
Atayal	84	.86	.82	.36
Cambodians	50	.69	.72	-.22
Nasioi	46	.66	.74	.37
Micronesians	74	.91	.86	.19
Yakut	102	.70	.70	.59
Cheyenne	112	.73	.71	.85
Arizona Pima	102	.85	.82	.72
Mexican Pima	106	.68	.65	.33
Maya	104	.42	.47	.72
Ticuna	132	.78	.78	.86
Rondonian Surui	92	.75	.70	.54
Karitiana	110	.95	.99	.42

NOTE.—Negative estimates of r are shown in bold.

^a N = number of individuals recruited in each study.

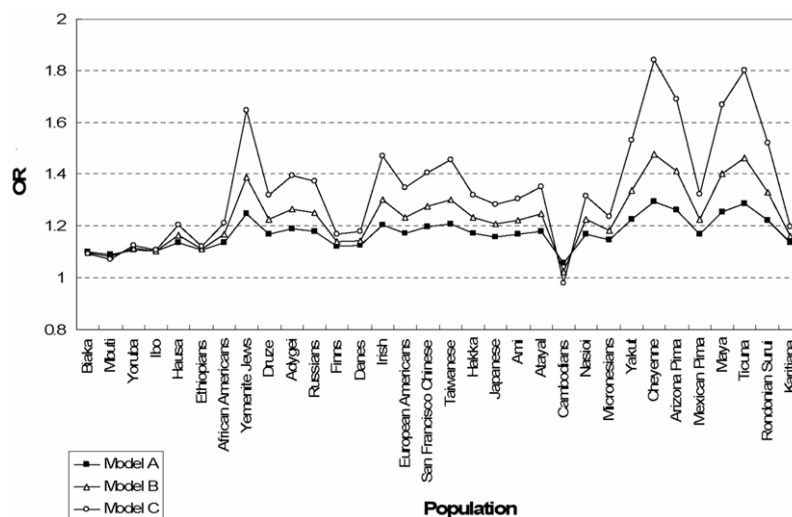


Figure 4. The effect of the Val allele (i.e., the A allele at *NlaIII* RSP) on SCZD across 32 populations. Allelic effect is presented by OR. For model A, $\theta_1 = 0.1$ and $\theta_2 = 0.2$; for model B, $\theta_1 = 0.1$ and $\theta_2 = 0.4$; for model C, $\theta_1 = 0.1$ and $\theta_2 = 0.8$. For all four models, $f = 0.001$.

and an *NlaIII* RFLP containing the Val158Met polymorphism in exon 4 of the *COMT* gene. Table 1 summarizes, for each population, the allele frequencies of A (site absent) and B (site absent) in the *NlaIII* and *HindIII* RFLPs in the *COMT* gene, respectively, and r between these two loci. We see that estimates of allele frequencies and allelic correlation vary across the populations. The effect of the A allele on disease risk was modeled under several putative disease models in these 32 populations (fig. 4). We found that the observed magnitude of association between the A allele and risk of disease varied across different populations. In the Cambodian population, allele A became overrepresented in unaffected individuals when the effect ratio—that is, $(1 + \theta_2)/(1 + \theta_1)$ —for the P2 promoter compared with the Val allele exceeds 1.4 under model I (fig. 5). Note that a previous study estimated ORs of 1.6 and 1.1 for the effects of the P2 promoter polymorphism and the Val allele, respectively, on SCZD, which is consistent with an effect ratio in excess of 1.4.²⁰

Part III: The *GAPDH* and *APOE* genes in AD.—The Alzheimer Disease Family data set and genotyping methods for *GAPDH* and *APOE* were described elsewhere.²¹ Briefly, the overall data set was composed of 653 white individuals: 369 unrelated subjects with AD and 284 unrelated control subjects. The mean (\pm SD) AAO for the affected subjects is 71.1 ± 6.6 , and controls were age matched at the time of examination (AAE), with a mean AAE of 71.9 ± 6.3 . Twelve SNPs in the *GAPDH* gene and its paralog were genotyped using the ABI 7900 TaqMan system. *APOE* alleles were genotyped as reported elsewhere.²²

Li et al.²³ reported that the G allele at a 5' UTR SNP—*rs3741916* (C→G)—in the *GAPDH* gene was positively associated with AD, whereas Lin et al.²¹ (from our laboratory)

reported that the C allele at the same locus was the positively associated allele in an independent AD sample. The original study reported that the G allele was positively associated with AD in the subset, without any copies of the *APOE* $\epsilon 4$ allele, whereas Lin et al.²¹ found that the same allele was inversely associated with AD in the subset carrying at least one copy of *APOE* $\epsilon 4$ alleles (a well-documented risk factor for AD). We thus sought to examine whether such opposite effects could be attributed to the variation in correlation between the *rs3741916* SNP and the *APOE* polymorphism.

Table 2 summarizes the findings of association between *rs3741916* in the *GAPDH* gene and risk of AD. We found that, under a recessive model, the G allele at *rs3741916* was positively associated with AD in the earlier-onset subset ($P = .007$) but that the C allele was positively associated with AD in the later-onset subset ($P = .047$). We fur-

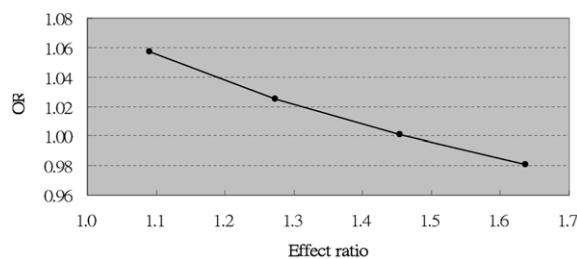


Figure 5. The effect of the Val allele (presented by OR) on SCZD in the Cambodian population, given different disease models (i.e., different effect ratios for the P2 promoter compared with the Val allele).

Table 2. Genotypic Relative Risk for *rs3741916* in the *GAPDH* Gene and the Correlation between *rs3741916* and the *APOE* Polymorphism

AAO	<i>rs3741916</i> and Disease		<i>rs3741916</i> and <i>APOE</i> Polymorphism		
	OR (95% CI)	<i>P</i>	OR ^a (95% CI)	<i>P</i>	<i>r</i> ^b
≤72 years (<i>N</i> = 195)	.36 ^c (.17–.76)	.007	.27 (.07–.99)	.049	–.021
>72 years (<i>N</i> = 211)	2.78 ^d (1.02–7.62)	.047	.92 (.32–2.70)	.884	.005

NOTE—The genotypic relative risk is presented as an OR.

^a The relative risk of carrying at least one *APOE* ε4 allele for individuals carrying at least one G allele at *rs3741916*, compared with individuals carrying no G alleles at *rs3741916*.

^b The *APOE* polymorphism was treated as a biallelic marker (the ε4 allele vs. the non-ε4 allele).

^c The genotype at *rs3741916* in the *GAPDH* gene was coded as 1 for CC and 0 for GC+CC.

^d The genotype at *rs3741916* in the *GAPDH* gene was coded as 1 for GG and 0 for GC+CC.

ther found that the G allele at *rs3741916* was inversely correlated with *APOE* ε4 allele ($r = -0.021$; $P = .049$) in the earlier-onset subset in our data set. On the contrary, the G allele at *rs3741916* was not significantly correlated with the *APOE* ε4 allele in the later-onset subset of our data set ($r = 0.005$; $P = .884$). The data set (of subjects recruited by Washington University in St. Louis) analyzed by Li and colleagues²³ included subjects with AD who had a mean AAO of 76.2 ± 6.8 years. The subjects included in our data set had significantly younger AAO than did the subjects included in the Washington University data set ($t = 11.1$; $P < .001$). These results suggest that opposite alleles are associated with AD in these two different samples because of differences in the interlocus correlation between *APOE* ε4 and *GAPDH* alleles.

The current study demonstrates that the observed effect of a genetic variant can differ between studies because of differences in its correlation with other causal variants. Association studies are often conducted on a marker-by-marker basis. For a complex disease, the interplay of multiple loci and environmental factors plays a role in its etiology. Therefore, examining the association between a single locus and disease may produce ambiguous results because of the lack of consideration of other disease-influencing genetic loci or environmental factors correlated with the target susceptibility locus.

Not only can flip-flop associations occur when a multilocus effect is not accounted for, but they may also be attributable to investigation of a noncausal variant in LD with a genuine causal variant. When the investigated allele at the noncausal variant is positively associated with the disease-risk allele in a population, the target allele at this SNP appears to be a risk allele. When this target allele at this SNP is negatively associated with the disease-risk allele in another population, the target allele at this SNP will appear to be a protective allele.

There are two scenarios we need to address separately. First, flip-flop phenomena may be attributable to genuinely different LD architectures. We have found notable difference in LD patterns in the *COMT* gene across populations with different ancestral origins. Therefore, flip-flop associations of the Val allele with SCZD across different ethnic

populations (e.g., Irish vs. Cambodian) could easily be due to the variation in LD architecture. Second, flip-flop phenomena may also be attributable to sampling variation that leads to variation in observed LD patterns. In this scenario, we may see flip-flop phenomena in different samples of the same ethnic origin, particularly when there is low LD between loci within the same gene.

LD patterns across the same genetic region can be highly variable across different ethnic populations.^{24,25} Variation in LD architecture across different populations within the same ethnic group has also been noted.²⁶ LD patterns are modulated by several factors, including local recombination rate,²⁷ mutation-selection balance, genetic drift, and population history. Another crucial factor, sampling variation, may lead to more-extreme swings in observed LD that may produce controversial opposite single-locus effects, as our theoretical modeling predicts. Therefore, we suggest that LD patterns be examined in different samples whenever flip-flop phenomena occur. If the samples show different patterns of LD between markers in the region, then this may explain the flip-flop. According to our theoretical modeling results, markers in weak LD with each other (e.g., $r^2 < 0.3$) may also need to be considered carefully, since flip-flop associations may be due only to sampling variation. Additionally, for loci with a minor-allele frequency <5%, it may be more likely that flip-flop phenomena indicate spurious results than confirmatory results confounded by the multilocus effect.

In summary, single-locus association studies may be complicated by interaction between an investigated locus and other risk factors. Consequently, conventional association tests performed in a “one-marker-at-a-time” fashion may produce ambiguous results, despite the genuine genetic effect. The current study shows that flip-flop associations may be partially attributable to population variation in interlocus correlation when the loci are not considered jointly. In-depth investigation is required to verify whether such phenomena indicate a confirmation or spurious association. These contradictory results should not be dismissed out of hand, and we should remember that genomic context is critical to the accurate interpretation of association results. Indeed, we would adapt the famous

work of John Donne²⁸: “No gene is an island, entire of itself, every gene is a piece of the interactive genome.”

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Web Resource

The URL for data presented herein is as follows:

Online Mendelian Inheritance in Man (OMIM), <http://www.ncbi.nlm.nih.gov/Omim/> (for autistic disorder, *GSTO1*, *AD*, *COMT*, *SCZD*, and *GAPDH*)

References

1. Devlin B, Cook EH Jr, Coon H, Dawson G, Grigorenko EL, McMahon W, Minshew N, Pauls D, Smith M, Spence MA, et al (2005) Autism and the serotonin transporter: the long and short of it. *Mol Psychiatry* 10:1110–1116
2. Betancur C, Corbex M, Spielewoy C, Philippe A, Laplanche JL, Launay JM, Gillberg C, Mouren-Simeoni MC, Hamon M, Giros B, et al (2002) Serotonin transporter gene polymorphisms and hyperserotonemia in autistic disorder. *Mol Psychiatry* 7:67–71
3. Maestrini E, Lai C, Marlow A, Matthews N, Wallace S, Bailey A, Cook EH, Weeks DE, Monaco AP, The International Molecular Genetic Study of Autism Consortium (1999) Serotonin transporter (*5-HTT*) and γ -aminobutyric acid receptor subunit $\beta 3$ (*GABRB3*) gene polymorphisms are not associated with autism in the IMGSA families. *Am J Med Genet* 88:492–496
4. McCauley JL, Olson LM, Dowd M, Amin T, Steele A, Blakely RD, Folstein SE, Haines JL, Sutcliffe JS (2004) Linkage and association analysis at the serotonin transporter (*SLC6A4*) locus in a rigid-compulsive subset of autism. *Am J Med Genet B Neuropsychiatr Genet* 127:104–112
5. Tordjman S, Gutknecht L, Carlier M, Spitz E, Antoine C, Slama F, Carsalade V, Cohen DJ, Ferrari P, Roubertoux PL, et al (2001) Role of the serotonin transporter gene in the behavioral expression of autism. *Mol Psychiatry* 6:434–439
6. Yirmiya N, Pilowsky T, Nemanov L, Arbelle S, Feinsilver T, Fried I, Ebstein RP (2001) Evidence for an association with the serotonin transporter promoter region polymorphism and autism. *Am J Med Genet* 105:381–386
7. Li YJ, Oliveira SA, Xu P, Martin ER, Stenger JE, Scherzer CR, Hauser MA, Scott WK, Small GW, Nance MA, et al (2003) Glutathione S-transferase omega-1 modifies age-at-onset of Alzheimer disease and Parkinson disease. *Hum Mol Genet* 12:3259–3267
8. Kolsch H, Linnebank M, Lutjohann D, Jessen F, Wullner U, Harbrecht U, Thelen KM, Kreis M, Hentschel F, Schulz A, et al (2004) Polymorphisms in glutathione S-transferase omega-1 and AD, vascular dementia, and stroke. *Neurology* 63:2255–2260
9. Chen X, Wang X, O’Neill AF, Walsh D, Kendler KS (2004) Variants in the catechol-O-methyltransferase (*COMT*) gene are associated with schizophrenia in Irish high-density families. *Mol Psychiatry* 9:962–967
10. Glatt SJ, Faraone SV, Tsuang MT (2003) Association between a functional catechol O-methyltransferase gene polymorphism and schizophrenia: meta-analysis of case-control and family-based studies. *Am J Psychiatry* 160:469–476
11. Park TW, Yoon KS, Kim JH, Park WY, Hirvonen A, Kang D (2002) Functional catechol-O-methyltransferase gene polymorphism and susceptibility to schizophrenia. *Eur Neuropsychopharmacol* 12:299–303
12. Sanders AR, Rusu I, Duan J, Vander Molen JE, Hou C, Schwab SG, Wildenauer DB, Martinez M, Gejman PV (2005) Haplotypic association spanning the 22q11.21 genes *COMT* and *ARVCF* with schizophrenia. *Mol Psychiatry* 10:353–365
13. Wonodi I, Stine OC, Mitchell BD, Buchanan RW, Thaker GK (2003) Association between Val108/158 Met polymorphism of the *COMT* gene and schizophrenia. *Am J Med Genet B Neuropsychiatr Genet* 120:47–50
14. Singleton AB, Farrer M, Johnson J, Singleton A, Hague S, Kachergus J, Hulihan M, Peuralinna T, Dutra A, Nussbaum R, et al (2003) Alpha-synuclein locus triplication causes Parkinson’s disease. *Science* 302:841
15. Tan EK, Tan C, Shen H, Chai A, Lum SY, Teoh ML, Yih Y, Wong MC, Zhao Y (2003) Alpha synuclein promoter and risk of Parkinson’s disease: microsatellite and allelic size variability. *Neurosci Lett* 336:70–72
16. Colhoun HM, McKeigue PM, Davey Smith G (2003) Problems of reporting genetic associations with complex outcomes. *Lancet* 361:865–872
17. Patsopoulos NA, Ntzani EE, Zintzaras E, Ioannidis JP (2005) *CYP2D6* polymorphisms and the risk of tardive dyskinesia in schizophrenia: a meta-analysis. *Pharmacogenet Genomics* 15:151–158
18. Marchini J, Donnelly P, Cardon LR (2005) Genome-wide strategies for detecting multiple loci that influence complex diseases. *Nat Genet* 37:413–417
19. Palmatier MA, Pakstis AJ, Speed W, Paschou P, Goldman D, Odunsi A, Okonofua F, Kajuna S, Karoma N, Kungulilo S, et al (2004) *COMT* haplotypes suggest P2 promoter region relevance for schizophrenia. *Mol Psychiatry* 9:859–870
20. Fan JB, Zhang CS, Gu NF, Li XW, Sun WW, Wang HY, Feng GY, St Clair D, He L (2005) Catechol-O-methyltransferase gene Val/Met functional polymorphism and risk of schizophrenia: a large-scale association study plus meta-analysis. *Biol Psychiatry* 57:139–144
21. Lin PI, Martin ER, Bronson PG, Browning-Large C, Small GW, Schmechel DE, Welsh-Bohmer KA, Haines JL, Gilbert JR, Pericak-Vance MA (2006) Exploring the association of glyceraldehyde-3-phosphate dehydrogenase gene and Alzheimer disease. *Neurology* 67:64–68
22. Saunders AM, Strittmatter WJ, Schmechel D, George-Hyslop PH, Pericak-Vance MA, Joo SH, Rosi BL, Gusella JF, Crapper-MacLachlan DR, Alberts MJ, et al (1993) Association of apolipoprotein E allele epsilon 4 with late-onset familial and sporadic Alzheimer’s disease. *Neurology* 43:1467–1472
23. Li Y, Nowotny P, Holmans P, Smemo S, Kauwe JS, Hinrichs AL, Tacey K, Doil L, van Luchene R, Garcia V, et al (2004) Association of late-onset Alzheimer’s disease with genetic variation in multiple members of the *GAPDH* gene family. *Proc Natl Acad Sci USA* 101:15688–15693
24. Schneider JA, Punliya MS, Choi JY, Jiang R, Sun XJ, Salisbury

- BA, Stephens JC (2003) DNA variability of human genes. *Mech Ageing Dev* 124:17–25
25. Nielsen R, Hubisz MJ, Clark AG (2004) Reconstituting the frequency spectrum of ascertained single-nucleotide polymorphism data. *Genetics* 168:2373–2382
26. Faraone SV, Skol AD, Tsuang DW, Bingham S, Young KA, Prabhudesai S, Haverstock SL, Mena F, Menon AS, Bisset D, et al (2002) Linkage of chromosome 13q32 to schizophrenia in a large veterans affairs cooperative study sample. *Am J Med Genet* 114:598–604
27. Mullerova J, Hozak P (2004) Use of recombinant congenic strains in mapping disease-modifying genes. *News Physiol Sci* 19:105–109
28. Donne J (1623) *Meditation XVII: devotions upon emergent occasions*. Oxford University Press, Oxford, United Kingdom (published in 1986)