



# Polymorphisms in Tumor Necrosis Factor- $\alpha$ Are Associated With Higher Anxiety Levels in Women After Breast Cancer Surgery

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## Abstract

**The present study identified 2 subgroups of patients with higher and lower levels of anxiety during the 6 months after breast cancer surgery. The women in the higher anxiety class were younger and reported a lower functional status. In addition, 2 polymorphisms in the tumor necrosis factor- $\alpha$  gene were associated with membership in the higher anxiety class.**

**Introduction:** Before and after breast cancer surgery, women have reported varying anxiety levels. Recent evidence has suggested that anxiety has a genetic basis and is associated with inflammation. The purposes of the present study were to identify the subgroups of women with distinct anxiety trajectories; to evaluate for differences in the phenotypic characteristics between these subgroups; and to evaluate for associations between polymorphisms in cytokine genes and subgroup membership. **Patients and Methods:** Patients with breast cancer ( $n = 398$ ) were recruited before surgery and followed up for 6 months. The patients completed the Spielberger State Anxiety Inventory and provided a blood sample for genomic analyses. Growth mixture modeling was used to identify the subgroups of patients with distinct anxiety trajectories. **Results:** Two distinct anxiety subgroups were identified. The women in the higher anxiety subgroup were younger and had a lower functional status score. Two single nucleotide polymorphisms in tumor necrosis factor- $\alpha$  ( $rs1799964$ ,  $rs3093662$ ) were associated with the higher anxiety subgroup. **Conclusion:** The results of the present exploratory study suggest that polymorphisms in cytokine genes could partially explain the interindividual variability in anxiety. The determination of phenotypic and molecular markers associated with greater levels of anxiety can assist clinicians to identify high-risk patients and initiate appropriate interventions.

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## Introduction

Moderate to high levels of psychological distress have been reported by most women before breast cancer surgery that gradually decreases during the first 12 months after surgery.<sup>1-12</sup> Across a number of studies, a younger age<sup>3,10,13</sup> and child care

responsibilities<sup>3</sup> were associated with greater levels of psychological distress. Of the clinical characteristics, although no differences in psychological distress were found between women who had undergone breast conserving surgery and those who had undergone mastectomy,<sup>4,9,10</sup> the receipt of adjuvant treatment was associated with greater levels of distress.<sup>3,14</sup> However, one of the major limitations of these studies was that “general measures” of psychological distress were used rather than a specific measure of anxiety (eg, Spielberger State-Trait Anxiety Inventory [STAI]<sup>15</sup>). Therefore, additional research is warranted to determine which demographic and clinical characteristics are specifically associated with greater levels of anxiety that persist after surgery.

Although the phenotypic characteristics that place patients with breast cancer at greater risk of more severe and persistent anxiety require additional investigation, the results from recent meta-analyses suggest that anxiety disorders have a genetic

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basis.<sup>16-21</sup> In addition, an emerging body of evidence suggests that inflammatory mechanisms may contribute to the development of anxiety disorders (for review, see Sokolowska and Hovatta<sup>22</sup> and Miller et al<sup>23</sup>). These investigators' reviews suggest that acute physical or psychological stressors modulate cytokine expression in the central nervous system. These changes in cytokine expression are adaptive, temporary, and controlled. However, when the stress becomes chronic, which is the case with a chronic illness such as cancer, the resultant chronic inflammatory responses contribute to the development of maladaptive behavioral symptoms (eg, anxiety, depression) and neuropsychiatric disorders.

Because of the associations between stress and its associated inflammatory responses and greater levels of symptoms,<sup>24,25</sup> our research team investigated the role of cytokine gene polymorphisms and an increased risk for depression,<sup>26</sup> pain,<sup>27,28</sup> sleep disturbances,<sup>29</sup> and attentional fatigue<sup>30</sup> in patients before and after surgery for breast cancer. In these studies, we used growth mixture modeling (GMM) to identify subgroups of patients with distinct symptom trajectories. In the present study, we extended this approach to an evaluation of state anxiety in women who were enrolled before breast cancer surgery and followed up for 6 months. The purposes of the present study, in a sample of 398 women who had undergone surgery for breast cancer, were to identify the subgroups (ie, latent classes) of women with distinct anxiety trajectories; to evaluate for differences in phenotypic characteristics between the latent classes; and to evaluate for associations between polymorphisms in genes for pro- and anti-inflammatory cytokines, their receptors, and their transcriptional regulators and latent class membership.

## Patients and Methods

### Patients and Settings

The present analysis was a part of a larger, longitudinal study that evaluated neuropathic pain and lymphedema in women who had undergone breast cancer surgery. The study methods have been previously described in detail.<sup>27,31-33</sup> In brief, the patients were recruited from breast care centers located at a comprehensive cancer center, 2 public hospitals, and 4 community practices.

Patients were eligible to participate if they were adult women (age  $\geq 18$  years) scheduled to undergo breast cancer surgery on 1 breast; who were able to read, write, and understand English; and who had agreed to participate and given written informed consent. Patients were excluded if they were scheduled to undergo bilateral breast cancer surgery or had distant metastases at the diagnosis. A total of 516 patients were approached, 410 were enrolled (response rate, 79.5%), and 398 completed the enrollment assessment. The most common reasons for refusal were too busy, overwhelmed with the cancer diagnosis, and insufficient time available to complete the enrollment assessment before surgery.

### Instruments

The demographic questionnaire obtained information on age, marital status, education, ethnicity, employment status, and living situation. Patients rated their functional status using the Karnofsky performance status (KPS) scale that ranged from 30 (I feel severely disabled and need to be hospitalized) to 100 (I feel normal; I have no complaints or symptoms).<sup>34</sup> The self-administered comorbidity

questionnaire (SCQ) was used to evaluate comorbidity.<sup>35</sup> Patients were asked to indicate whether they had 1 of 13 common medical conditions; whether they had received treatment for it (proxy for disease severity); and whether it had limited their activities (indication of functional limitations). For each condition, a patient could receive a maximum of 3 points, with a maximum total score of 39. The SCQ has well-established validity and reliability.<sup>36,37</sup>

The Spielberger State-Trait Anxiety Inventories (STAI-T, STAI-S) consist of 20 items, each rated from 1 to 4. The scores for each scale are summed and range from 20 to 80. A higher score indicates greater anxiety. The STAI-T measures an individual's predisposition to anxiety determined by the person's personality and estimates how a person generally feels. The STAI-S measures an individual's transitory emotional response to a stressful situation. It evaluates the emotional responses of worry, nervousness, tension, and feelings of apprehension related to how a person feels "right now" in a stressful situation. A cutoff score of  $\geq 31.8$  and  $\geq 32.2$  indicates a high level of trait and state anxiety, respectively.<sup>15</sup> The STAI-S and STAI-T inventories have well-established validity and reliability.<sup>38,39</sup> In the present study, Cronbach's  $\alpha$  for the STAI-T and STAI-S were 0.88 and 0.95, respectively.

The Center for Epidemiological Studies Depression Scale (CES-D) consists of 20 items selected to represent the major symptoms in the clinical syndrome of depression. The scores range from 0 to 60, with scores of  $\geq 16$  indicating the need for individuals to seek clinical evaluation for major depression. The CES-D has well-established validity and reliability.<sup>40-42</sup> In the present study, Cronbach's  $\alpha$  for the CES-D was 0.90.

### Study Procedures

The Committee on Human Research at the University of California, San Francisco, and the institutional review boards at each of the study sites approved the present study. During the patient's preoperative visit, a clinician explained the study, determined the patient's willingness to participate, and introduced the patient to the research nurse. The research nurse met with the women, determined their eligibility, and obtained written informed consent before surgery. After obtaining consent, the patients completed the enrollment questionnaires an average of 4 days before their surgery. The patients completed the STAI-S at enrollment and monthly for 6 months (ie, 7 assessments). The patients' medical records were reviewed for disease and treatment information.

### Genomic Analyses

**Gene Selection.** Cytokines, their receptors, and their transcriptional regulators are classes of polypeptides that mediate inflammatory processes. Cytokine dysregulation has been associated with anxiety (for review, see Sokolowska and Hovatta<sup>22</sup> and Miller et al<sup>23</sup>). These polypeptides are divided into pro- and anti-inflammatory cytokines. Pro-inflammatory mediators promote systemic inflammation and include interferon- $\gamma$  (IFNG), IFNG receptor 1 (IFNGR1), interleukin-1 receptor 1 (IL1R1), interleukin (IL) 2, IL8, IL17A, and tumor necrosis factor- $\alpha$  (TNFA). Anti-inflammatory mediators suppress the activity of pro-inflammatory cytokines and include IL1R2, IL4, IL10, and IL13. Of note, IFNGR1, IL1B, and IL6 possess pro- and anti-inflammatory

functions. Nuclear factor  $\kappa\beta$ -1 (NFKB1) and NFKB2 are transcriptional regulators of these cytokine genes.<sup>43</sup>

**Blood Collection and Genotyping.** Of the 398 patients who completed the baseline assessment, 310 provided a blood sample from which DNA could be isolated from the peripheral blood mononuclear cells (PBMCs). Genomic DNA was extracted from PBMCs using the PUREGene DNA Isolation System (Invitrogen, Carlsbad, CA). DNA was quantified using a Nanodrop spectrophotometer (model no. ND-1000; Thermo Scientific, ThermoFisher Scientific, Waltham, MA) and normalized to a concentration of 50 ng/ $\mu$ L. Genotyping was performed without awareness of the patients' clinical status, and positive and negative controls were included. The samples were genotyped using the Golden Gate genotyping platform (Illumina, San Diego, CA) and processed according to the standard protocol using Genome Studio (Illumina). Two reviewers who were unaware of the patient characteristics visually inspected the signal intensity profiles and resulting genotype calls for each single nucleotide polymorphism (SNP).

**SNP Selection.** A combination of tagging SNPs and literature-driven SNPs were selected for analysis. The tagging SNPs were required to be common (ie, a minor allele frequency  $\geq 0.05$ ) in public databases. SNPs with call rates of  $< 95\%$  or Hardy-Weinberg  $P$  values of  $< .001$  were excluded. As listed in Supplemental Table 1 (in the online version), a total of 82 SNPs among the 15 candidate genes passed all the quality control filters and were included in the genetic association analyses. The potential functional roles of the SNPs associated with state anxiety were examined using PUPASuite, version 2.0.<sup>44</sup>

### Statistical Analyses for Phenotypic Data

The data were analyzed using the Statistical Package for Social Sciences,<sup>45</sup> version 20 (IBM Corp, Armonk, NY), and STATA,<sup>46</sup> version 13 (StataCorp, College Station, TX).<sup>46</sup> Descriptive statistics and frequency distributions were generated for sample characteristics. Independent sample  $t$  tests, Mann-Whitney  $U$  tests, and  $\chi^2$  analyses were used to evaluate for differences in the demographic and clinical characteristics between the 2 latent classes. All calculations used actual values. Adjustments were not made for missing data. Therefore, the cohort for each analysis was dependent on the largest set of available data between the 2 groups.

Unconditional GMM with robust maximum likelihood estimation was performed to identify latent classes with distinct anxiety trajectories using *Mplus*, version 5.21 (available at: [www.statmodel.com](http://www.statmodel.com)). These methods have been previously described in detail.<sup>33</sup> In brief, a single growth curve that represented the "average" change trajectory was estimated for the whole sample. Next, the number of latent growth classes that best fit the data was identified using the guidelines recommended in published studies.<sup>47-49</sup>

### Statistical Analyses for Genetic Data

The allele and genotype frequencies were determined by gene counting. Hardy-Weinberg equilibrium was assessed using the  $\chi^2$  or

Fisher exact test. Measures of linkage disequilibrium (LD; ie,  $D'$  and  $r^2$ ) were computed from the patients' genotypes using Haploview, version 4.2 (Broadview Institute, Cambridge, MA). The LD-based haplotype block definition was based on the  $D'$  confidence interval.

For the SNPs that were members of the same haploblock, haplotype analyses were conducted to localize the association signal within each gene and to determine whether the haplotypes improved the strength of the association with the phenotype. Haplotypes were constructed using the program PHASE, version 2.1.<sup>50</sup> To improve the stability of haplotype inference, the haplotype construction procedure was repeated 5 times using different seed numbers with each cycle. Only those haplotypes that were inferred with probability estimates of  $\geq .85$ , across the 5 iterations, were retained for the downstream analyses. Only the inferred haplotypes that occurred with a frequency estimate of  $\geq 15\%$  were included in the association analyses, assuming a dosage model (ie, analogous to the additive model).

Ancestry informative markers (AIMs) were used to minimize confounding due to population stratification.<sup>51-53</sup> The homogeneity in ancestry among the patients was verified by principal component analysis<sup>54</sup> using HelixTree (GoldenHelix, Bozeman, MT). In brief, the number of principal components (PCs) was sought that distinguished the major racial or ethnic groups in the sample by visual inspection of the scatter plots of the orthogonal PCs (ie, PC1 vs. PC2, PC2 vs. PC3). This procedure was repeated until no discernable clustering of patients by their self-reported race/ethnicity was possible (data not shown). The first 3 PCs were selected to adjust for potential confounding due to population substructure (ie, race/ethnicity) by including them in all logistic regression models. Finally, 106 AIMs were included in the analysis.

For the association tests, 3 genetic models were assessed for each SNP: additive, dominant, and recessive. Barring trivial improvements (ie, change  $< 10\%$ ), the genetic model that best fit the data, by maximizing the significance of the  $P$  value, was selected for each SNP. Logistic regression analysis, controlling for significant covariates and genomic estimates of and self-reported race/ethnicity, were used to evaluate the association between genotype and anxiety class. Only those genetic associations identified as significant from the bivariate analyses were evaluated in the multivariate analyses. A backward stepwise approach was used to create a parsimonious model. Except for the genomic estimates of and self-reported race/ethnicity, only the predictors with  $P < .05$  were retained in the final model. The genetic model fit and both unadjusted and covariate-adjusted odds ratios were estimated using STATA, version 13 (StataCorp, College Station, TX).<sup>46</sup>

Just as was performed in our previous studies,<sup>55-57</sup> and in accordance with the recommendations in published studies,<sup>58,59</sup> with the implementation of rigorous quality controls for genomic data, the nonindependence of SNPs/haplotypes in LD, and the exploratory nature of the analyses, adjustments were not made for multiple testing. In addition, the significant SNPs identified in the bivariate analyses were evaluated further using logistic regression analyses that controlled for differences in phenotypic characteristics, potential confounding due to population stratification, and variations in other SNPs/haplotypes within the same gene. Only those SNPs that remained significant were included in the final presentation of the results. Therefore, the significant independent associations reported

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were unlikely to have resulted from chance. Unadjusted (bivariate) associations are reported for all the SNPs that passed the quality control criteria in [Supplemental Table 1](#) (available in the online version), to allow for subsequent comparisons and meta-analyses.

## Results

### GMM Analysis

Two distinct latent classes of anxiety trajectories were identified using GMM ([Figure 1](#)). A 2-class model was selected because its Bayesian information criterion was smaller than that of the 1-class model. In addition, a 3-class solution could not be reliably fit ([Table 1](#)).

As listed in [Table 2](#), most patients were classified into the higher anxiety group (63.1%). These patients had state anxiety scores that were high at enrollment ( $47.6 \pm 12.4$ ) and that gradually decreased during the 6 months of the study. The patients in the lower anxiety group (36.9%) had state anxiety scores that were lower at enrollment ( $31.8 \pm 8.8$ ) and that gradually decreased over time.

### Differences in Demographic and Clinical Characteristics

As summarized in [Table 3](#), the patients in the higher anxiety class were significantly younger and had a lower KPS score and a higher SCQ score. In addition, a greater percentage of patients in the higher anxiety class had a Hispanic/mixed ethnic background (compared with whites), had received neoadjuvant chemotherapy (CTX), and had received adjuvant CTX during the first 6 months after breast cancer surgery. The patients in the higher anxiety class had higher trait anxiety scores and higher CES-D scores at enrollment.

### Candidate Gene Analyses of Both GMM Classes

As summarized in [Supplemental Table 1](#) (available in the online version), the minor allele frequency was significantly different between the 2 latent classes for 6 SNPs and 2 haplotypes: IL1R1

**Table 1** Fit Indexes for the Spielberger State Anxiety Scale GMM Class Solutions for 398 Patients With Breast Cancer

GMM	LL	AIC	BIC	Entropy	BLRT	VLMR
1-Class <sup>a</sup>	-9174.21	18380.42	18444.20	NA	NA	NA
2-Class <sup>b</sup>	-9136.32	18314.64	18398.36	.63	75.78 <sup>c</sup>	75.78 <sup>d</sup>
3-Class <sup>e</sup>						

Abbreviations: AIC = Akaike information criterion; BIC = Bayesian information criterion; BLRT = parametric bootstrapped likelihood ratio test for K-1 (H0) versus K classes; GMM = growth mixture model; LL = log likelihood; VLMR = Vuong-Lo-Mendell-Rubin likelihood ratio test for K-1 (H0) versus K classes.

<sup>a</sup>Latent growth curve with linear and quadratic components:  $\chi^2 = 40.19$ ,  $df = 19$ ,  $P = .01$ , comparative fit index = 0.976, root mean square error of approximation = 0.053.

<sup>b</sup>A 2-class model was selected; the BIC was smaller than for the 1-class model, and the BLRT indicated that the 2-class solution fit the data better than did the 1-class solution.

<sup>c</sup> $P < .00005$ .

<sup>d</sup> $P < .01$ .

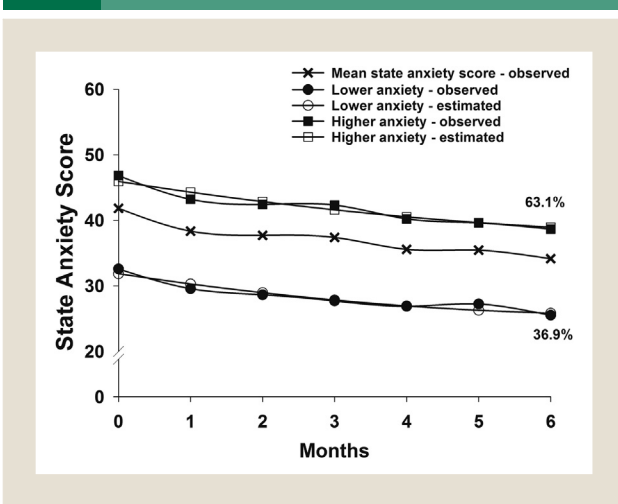
<sup>e</sup>A 3-class solution could not be reliably fit without fixing multiple ( $n = 29$ ) parameter estimates at 0, owing to singularity in the information matrix and nonpositive definite covariance matrices of the parameter estimates for 2 classes; even with the parameters fixed at 0, the smallest class was only 3% of the total, which is not a reliable size for a latent class.

rs3917332, IL6 rs2069840, IL6 HapA5, IL13 rs1295686, IL13 HapA1, NFKB2 rs1056890, TNFA rs1799964, and TNFA rs3093662.

### Regression Analyses for IL1R1, IL6, IL13, NFKB2, and TNFA Genotypes and Lower Versus Higher Anxiety Classes

To better estimate the magnitude (ie, odds ratio) and precision (95% confidence interval) of the genotype on the odds of belonging to the higher versus lower anxiety class, multivariate logistic regression models were fit. In these regression analyses, which

**Figure 1** Observed and Estimated State Anxiety Inventory Trajectories for Patients in Each of the Latent Classes and the Mean State Anxiety Scores for the Total Sample



**Table 2** Parameter Estimates for Spielberger State Anxiety Scale GMM Latent Classes From 7 Assessments of 398 Patients With Breast Cancer

Variable	Lower Anxiety Class <sup>a</sup> (n = 147)	Higher Anxiety Class <sup>a</sup> (n = 251)
Parameter estimates		
Intercept	31.85 $\pm$ 1.54 <sup>b</sup>	45.93 $\pm$ 0.89 <sup>b</sup>
Linear slope	-1.68 $\pm$ 0.62 <sup>c</sup>	-1.72 $\pm$ 0.57 <sup>c</sup>
Quadratic slope	0.11 $\pm$ 0.08	0.09 $\pm$ 0.08
Variances		
Intercept	10.52 $\pm$ 3.49 <sup>c</sup>	77.68 $\pm$ 13.24 <sup>b</sup>
Linear slope	0 <sup>d</sup>	29.71 $\pm$ 5.79 <sup>b</sup>
Quadratic slope	0 <sup>d</sup>	0.66 $\pm$ 0.14 <sup>b</sup>
I with S	0 <sup>d</sup>	-26.55 $\pm$ 8.03 <sup>b</sup>
I with Q	0 <sup>d</sup>	3.27 $\pm$ 1.15 <sup>c</sup>
S with Q	0 <sup>d</sup>	-4.20 $\pm$ 0.85 <sup>b</sup>

Data presented as mean  $\pm$  standard error.

Abbreviations: GMM = growth mixture model; I = intercept; Q = quadratic slope; S = linear slope.

<sup>a</sup>Predicted class size according to most likely class membership.

<sup>b</sup> $P < .001$ .

<sup>c</sup> $P < .01$ .

<sup>d</sup>Random intercepts model only; random slopes were fixed at 0 to assist in estimation.

**Table 3** Differences in Demographic and Clinical Characteristics Between Lower (n = 147) and Higher (n = 251) Latent Classes

Characteristic	Lower Anxiety Class (n = 147; 36.9%)	Higher Anxiety Class (n = 251; 63.1%)	Statistic and P Value
Age (years)	57.5 ± 11.5	53.4 ± 11.3	t = 3.46, P = .001
Education (years)	15.9 ± 2.6	15.6 ± 2.7	NS
KPS score	95.1 ± 9.5	92.1 ± 10.6	t = 2.80, P = .005
SCQ score	3.9 ± 2.4	4.5 ± 3.0	t = -2.07, P = .039
Trait anxiety score at enrollment	29.9 ± 5.6	38.5 ± 9.1	t = -11.3, P < .0001
State anxiety score at enrollment	31.8 ± 8.8	47.6 ± 12.4	t = -14.6, P < .0001
CES-D scale score at enrollment	7.6 ± 6.2	17.3 ± 9.7	t = -11.9, P < .0001
Breast biopsies in previous year (n)	1.4 ± 0.6	1.6 ± 0.9	NS
Positive lymph nodes (n)	0.7 ± 1.7	1.0 ± 2.5	NS
Lymph nodes removed (n)	4.8 ± 5.7	6.3 ± 7.2	NS
Ethnicity			$\chi^2 = 17.801$ , P < .0001; P = .008 <sup>a</sup>
White	76.7 (112)	57.2 (143)	
Black	7.5 (11)	11.6 (29)	
Asian/Pacific Islander	10.3 (15)	14.0 (35)	
Hispanic/mixed ethnic background/other	5.5 (8)	17.2 (43)	
Married/partnered (% yes)	36.7 (54)	44.9 (111)	NS
Work for pay (% yes)	50.3 (74)	46.4 (115)	NS
Lives alone (% yes)	21.8 (32)	25.6 (63)	NS
Postmenopausal (% yes)	67.1 (98)	62.2 (150)	NS
Disease stage			NS
0	20.4 (30)	17.1 (43)	
I	43.5 (64)	34.7 (87)	
IIA and IIB	31.3 (46)	37.8 (95)	
IIIA, IIIB, IIIC, IV	4.8 (7)	10.4 (26)	
Surgical treatment			NS
Breast conservation	80.3 (118)	79.7 (200)	
Mastectomy	19.7 (29)	20.3 (51)	
Sentinel node biopsy (% yes)	85.7 (126)	80.5 (202)	NS
Axillary lymph node dissection (% yes)	32.2 (47)	40.6 (102)	NS
Breast reconstruction at surgery (% yes)	22.6 (33)	21.1 (53)	NS
Neoadjuvant chemotherapy (% yes)	13.0 (19)	23.9 (60)	FE, P = .009
Radiotherapy during first 6 mo (% yes)	57.8 (85)	55.4 (139)	NS
Chemotherapy during first 6 mo (% yes)	25.9 (38)	37.8 (95)	FE, P = .016

Data presented as mean ± standard deviation or % (n).

Abbreviations: CES-D = Center for Epidemiological Studies depression scale; FE = Fisher's exact test; KPS = Karnofsky performance status; NS = not significant; SCQ = self-administered comorbidity questionnaire.

<sup>a</sup>White versus Hispanic/mixed/other.

included genomic estimates of and self-reported race/ethnicity, the only phenotypic characteristic that remained significant in the multivariate model was age (in 5-year increments).

The only genetic associations that remained significant in the multivariate logistic regression analyses were for TNFA rs1799964 and TNFA rs3093662 (Table 4, Figure 2). In the regression analysis for TNFA rs1799964, controlling for age and rs3093662, carrying 2 of the rare C allele (ie, TT+TC vs. CC) was associated with an 88% reduction in the odds of belonging to the higher anxiety class. In the same analysis, controlling for age and rs1799964, carrying 1 or 2 doses of the rare G allele (ie, AA vs. AG+GG) in TNFA rs3093662 was associated with a

4.04 increase in the odds of belonging to the higher anxiety class.

## Discussion

The present study is the first to use GMM to identify subgroups of women with distinct trajectories of anxiety from before through 6 months after breast cancer surgery. Before surgery, the women in the lower anxiety class reported both state and trait anxiety scores that approached the clinically meaningful cutoff score.<sup>15</sup> In contrast, in the higher anxiety class, both trait and state anxiety scores were greater than the cutoff scores. In the higher anxiety class, the state anxiety score of 47.6 was greater than the scores reported by patients

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**Table 4** Multiple Logistic Regression Analyses for State Anxiety and Candidate Gene Markers

Predictor	Odds Ratio	Standard Error	95% CI	Z	P Value
TNFA rs1799964	0.12	0.084	0.030-0.471	-3.03	.002
TNFA rs3093662	4.04	1.789	1.694-9.623	3.15	.002
Age	0.83	0.047	0.745-0.930	-3.26	.001

Multiple logistic regression analysis of GMM latent classes for state anxiety scores (0 = lower, 1 = higher). For each model, the first 3 principle components identified from analysis of ancestry informative markers and self-reported race/ethnicity were retained in all models to adjust for potential confounding due to race or ethnicity (data not shown). The predictors evaluated in each model included genotype (TNFA rs1799964 genotype: TT+TC vs. CC; and TNFA rs3093662 genotype: AA vs. AG+GG), age (in 5-year increments), and self-reported race/ethnicity (white [reference group], Asian/Pacific Islander, black, Hispanic/mixed/other). Overall model fit:  $\chi^2 = 42.81$ ,  $P < .0001$ . Abbreviations: CI = confidence interval; GMM = growth mixture model; TNFA = tumor necrosis factor- $\alpha$ .

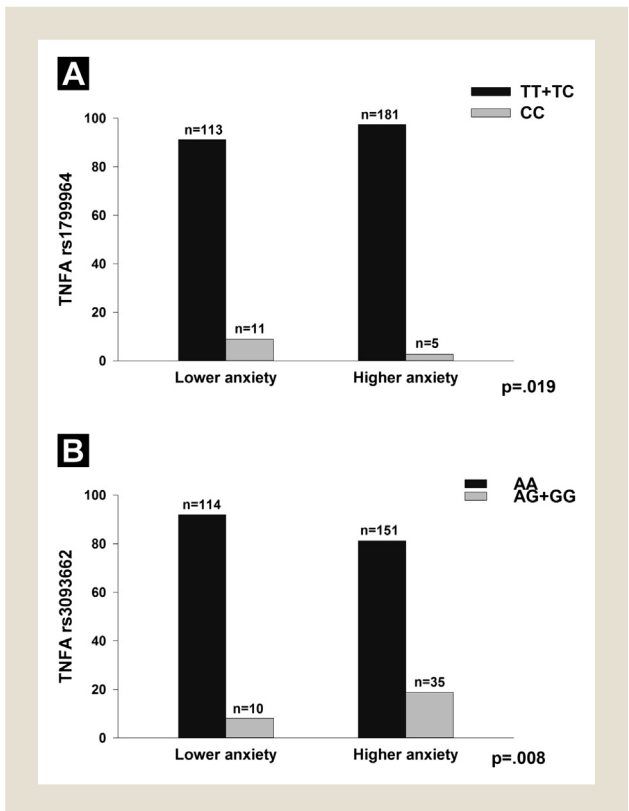
with coronary artery disease<sup>60</sup> (ie, 39.0), chronic renal failure<sup>61</sup> (ie, 28.7), or chronic obstructive pulmonary disease<sup>62</sup> (ie, 34.9) but somewhat lower than the scores reported by patients with generalized anxiety disorder<sup>63</sup> (ie, 52.2).

The differences in the preoperative trait and state anxiety scores between the lower and higher anxiety classes were not only statistically significant but clinically meaningful for both scores (ie,  $d = 1.0$  and  $d = 1.2$ , respectively, where  $d$  represents the difference between the 2 groups in standard deviation units).<sup>64,65</sup> Although the results of the GMM analysis demonstrated that in both anxiety classes, the severity of state anxiety decreased during the 6 months of the study, the reductions from before through 6 months after surgery were relatively modest (ie, 25.5 for the lower and 38.7 for the higher anxiety classes at 6 months). This finding suggests that for > 60% of the sample, relatively high levels of anxiety persisted after breast cancer surgery.

Age and ethnicity were the 2 demographic characteristics associated with the higher anxiety class. For each 5-year increase in age, the state anxiety scores decreased by approximately 1 point. Consistent with previous reports,<sup>3,10,13</sup> younger patients reported higher anxiety scores. In addition, compared with whites, the patients who self-reported their ethnicity as Hispanic or of mixed ethnic background were more likely to be included in the higher anxiety group. This finding is consistent with a recent meta-analysis that found that Hispanic oncology patients in the United States reported greater levels of psychological distress and poorer quality of life outcomes.<sup>66</sup>

A number of clinical and symptom characteristics were associated with the higher anxiety class. Consistent with previous findings,<sup>67-69</sup> patients in the higher anxiety class reported significantly lower KPS scores. In addition, and consistent with a previous report of oncology patients and their family caregivers,<sup>69</sup> patients with a more severe comorbidity profile were classified in the higher anxiety class. Although receipt of neoadjuvant CTX has not been identified as a risk factor for greater levels of anxiety, receipt of adjuvant CTX after breast cancer surgery was associated with greater levels of anxiety.<sup>3,14</sup> Finally, patients in the higher anxiety group reported depressive symptom scores that were greater than the clinically meaningful

**Figure 2** (A) Differences Between the Latent Classes in the Percentages of Patients Who Were Homozygous or Heterozygous for the Common Allele (TT + TC) or Homozygous for the Rare Allele (CC) for rs1799964 in Tumor Necrosis Factor- $\alpha$  (TNFA). Values Were Plotted as Unadjusted Proportions With the Corresponding P Value. (B) Differences Between the Latent Classes in the Percentages of Patients Who Were Homozygous for the Common Allele (AA) or Heterozygous or Homozygous for the Rare Allele (AG + GG) for rs3093662 in TNFA. Values Were Plotted as Unadjusted Proportions With Corresponding P Values



cutoff for the CES-D. In the 3 studies that evaluated the co-occurrence of anxiety and depressive symptoms (CAD) in patients with breast cancer,<sup>70-72</sup> the prevalence rates for CAD ranged from 10% to 28%, depending on the timing of the assessment. Taken together, clinicians can use these demographic, clinical, and symptom characteristics to identify patients who are at increased risk for clinically meaningful levels of anxiety before and after breast cancer surgery.

In the present study, 2 SNPs in TNFA (ie, rs1799964 and rs3093662) were associated with the higher anxiety class. Patients who were homozygous for the rare C allele in TNFA rs1799964 had an 88% decrease in the odds of belonging to the higher anxiety class. Although TNFA rs1799964 is 1 of the 8 SNPs used to infer HapA, rs3093662 is not included in this haplotype. Therefore, the finding that each SNP was independently associated with membership in the higher anxiety class is in agreement with the null association between the TNFA HapA and latent class membership. Although no studies were identified that evaluated an association

between this SNP and anxiety, several published findings are worth noting. In 1 study of patients with lung cancer,<sup>73</sup> this same SNP, which is located in the promoter region of the TNFA gene, was associated with decreased pain. In addition, in a community-based cohort study,<sup>74</sup> those who were homozygous for the rare C allele had a 17% decrease in the odds of cancer-related mortality. Finally, in the same sample of patients with breast cancer,<sup>26</sup> being homozygous for the rare C allele was associated with an 87% decrease in the odds of belonging to the subsyndromal depressive symptoms' class.

In contrast, patients who were heterozygous or homozygous for the rare G allele at rs3093662 were 4.0 times more likely to be in the higher anxiety class. Although this SNP lies in the intronic region of the TNFA gene, it was associated with inflammation and poorer outcomes in patients who had undergone renal transplantation.<sup>75</sup> In addition, in our previous study of patients who had undergone radiation therapy and their family caregivers,<sup>76</sup> the participants who were heterozygous or homozygous for the rare G allele were 3.8 times more likely to be classified in the higher morning fatigue class.

A growing body of evidence suggests that TNFA exerts a broad range of biologic functions within the peripheral and central nervous systems.<sup>77-80</sup> The findings from the present study suggest that polymorphisms in TNFA are associated with anxiety and other common symptoms in oncology patients. Although functional studies are needed to confirm these associations, our findings have been supported by a recent study that found that the administration of TNFA antagonists was associated with a decrease in the occurrence of generalized anxiety disorder in patients with rheumatoid arthritis.<sup>81</sup>

The present study had a number of study limitations. Although our sample size was adequate, larger, independent samples are needed to confirm these preliminary findings and identify additional latent classes, significant phenotypic predictors, and significant genetic associations. Also, although a valid and reliable self-report measure was used to evaluate state anxiety, future studies should incorporate a clinical evaluation of pre-existing and concurrent psychiatric conditions. In addition, no information was available on whether these patients were taking anti-anxiety medications. Finally, the generalizability of the study findings is limited to only female patients with breast cancer.

## Conclusion

Despite these limitations, these findings provide evidence to support distinct anxiety phenotypes in patients with breast cancer before and after surgery. In addition, in this sample, the higher risk phenotype was associated with higher levels of depressive symptoms before surgery. It is important that these higher risk patients be identified early to be able to provide pre-emptive and ongoing treatment.

## Clinical Practice Points

- Moderate to high levels of psychological distress are reported by most women after breast cancer surgery.
- Although the exact prevalence of anxiety is unknown in women after breast cancer surgery, recent evidence suggests that

inflammatory mechanisms might contribute to the development of anxiety disorders.

- Using GMM, 2 subgroups of women with distinct trajectories of anxiety (ie, higher and lower anxiety classes) were identified.
- Younger patients, those with Hispanic or mixed ethnic background, and those with lower functional status scores were more likely to be in the higher anxiety class.
- Two single nucleotide polymorphisms in TNFA (ie, rs1799964 and rs3093662) were associated with membership in the higher anxiety class.
- Clinicians can use the phenotypic predictors associated with higher anxiety to identify women who warrant additional evaluation before breast cancer surgery.

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## Disclosure

The authors have stated that they have no conflicts of interest.

## Supplemental Data

Supplemental tables accompanying this article can be found in the online version at <http://dx.doi.org/10.1016/j.clbc.2014.12.001>.

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**Supplemental Table 1** Summary of Cytokine Gene SNPs and Haplotypes Analyzed for Lower Versus Higher Anxiety Latent Classes

Gene	SNP	Position	Chromosome	Haplotype <sup>a</sup>	MAF	Alleles	$\chi^2$	P Value	Model
IFNG1	rs2069728	66834051	12	HapA	.110	G>A	3.460	.177	A
IFNG1	rs2069727	66834490	12	HapA	.384	A>G	1.928	.381	A
IFNG1	rs2069718	66836429	12	HapA	.494	C>T	0.119	.942	A
IFNG1	rs1861493	66837463	12	HapA	.266	A>G	0.545	.761	A
IFNG1	rs1861494	66837676	12	HapA	.273	T>C	0.719	.698	A
IFNG1	rs2069709	66839970	12		.003	G>T	NA	NA	NA
IFNG1	HapA3						0.467	.792	
IFNG1	HapA5						2.484	.289	
IFNGR1	rs9376268	137574444	6		.254	G>A	0.938	.626	A
IL1B	rs1071676	106042060	2	HapA	.189	G>C	1.669	.434	A
IL1B	rs1143643	106042929	2	HapA	.383	G>A	1.948	.378	A
IL1B	rs1143642	106043180	2	HapA	.082	C>T	0.171	.918	A
IL1B	rs1143634	106045017	2	HapA	.187	C>T	1.514	.469	A
IL1B	rs1143633	106045094	2	HapA	.392	G>A	2.469	.291	A
IL1B	rs1143630	106046282	2	HapB	.115	C>A	1.653	.438	A
IL1B	rs3917356	106046990	2	HapB	.450	G>A	1.076	.584	A
IL1B	rs1143629	106048145	2	HapB	.389	T>C	1.060	.589	A
IL1B	rs1143627	106049014	2	HapB	.397	T>C	0.636	.728	A
IL1B	rs16944	106049494	2	HapB	.386	G>A	0.394	.821	A
IL1B	rs1143623	106050452	2	HapB	.277	G>C	0.072	.965	A
IL1B	rs13032029	106055022	2	HapB	.448	C>T	0.784	.676	A
IL1B	HapA1						0.811	.667	
IL1B	HapA4						2.010	.366	
IL1B	HapA6						1.549	.461	
IL1B	HapB1						2.410	.300	
IL1B	HapB6						1.015	.602	
IL1B	HapB8						0.717	.699	
IL1R1	rs949963	96533648	2		.223	G>A	1.589	.452	A
IL1R1	rs2228139	96545511	2		.053	C>G	3.171	.205	A
IL1R1	rs3917320	96556738	2		.047	A>C	NA	NA	NA
IL1R1	rs2110726	96558145	2	HapA	.317	C>T	4.116	.128	A
IL1R1	rs3917332	96560387	2	HapA	.187	A>T	FE	.020	D
IL1R1	HapA1						3.231	.199	
IL1R1	HapA2						3.604	.165	
IL1R1	HapA3						5.971	.051	
IL1R2	rs4141134	96370336	2		.362	T>C	3.443	.179	A
IL1R2	rs11674595	96374804	2	HapA	.258	T>C	0.681	.711	A
IL1R2	rs7570441	96380807	2	HapA	.408	G>A	2.116	.347	A
IL1R2	HapA1						2.301	.316	
IL1R2	HapA2						FE	.117	
IL1R2	HapA4						0.050	.975	
IL2	rs1479923	119096993	4		.308	C>T	2.763	.251	A
IL2	rs2069776	119098582	4		.184	T>C	NA	NA	NA
IL2	rs2069772	119099739	4	HapA	.241	A>G	3.192	.203	A
IL2	rs2069777	119103043	4		.047	C>T	NA	NA	NA
IL2	rs2069763	119104088	4	HapA	.277	T>G	2.574	.276	A
IL2	HapA1						2.008	.366	
IL2	HapA2						2.703	.259	
IL2	HapA3						3.192	.203	
IL4	rs2243248	127200946	5		.086	T>G	3.239	.198	A
IL4	rs2243250	127201455	5	HapA	.269	C>T	NA	NA	NA
IL4	rs2070874	127202011	5	HapA	.245	C>T	NA	NA	NA

Supplemental Table 1 Continued

Gene	SNP	Position	Chromosome	Haplotype <sup>a</sup>	MAF	Alleles	$\chi^2$	P Value	Model
IL4	rs2227284	127205027	5	HapA	.387	C>A	NA	NA	NA
IL4	rs2227282	127205481	5	HapA	.390	C>G	NA	NA	NA
IL4	rs2243263	127205601	5	HapA	.124	C>G	0.181	.913	A
IL4	rs2243266	127206091	5	HapA	.237	G>A	NA	NA	NA
IL4	rs2243267	127206188	5	HapA	.237	G>C	NA	NA	NA
IL4	rs2243274	127207134	5	HapA	.261	G>A	NA	NA	NA
IL4	HapA1						0.658	.720	
IL4	HapA3						0.174	.917	
IL4	HapX1						3.396	.183	
IL6	rs4719714	22643793	7		.255	A>T	4.170	.124	A
IL6	rs2069827	22648536	7		.069	G>T	0.803	.669	A
IL6	rs1800796	22649326	7		.134	C>G	NA	NA	NA
IL6	rs1800795	22649725	7	HapA	.285	C>G	2.818	.244	A
IL6	rs2069830	22650951	7	HapA	.061	T>C	NA	NA	NA
IL6	rs2066992	22651329	7		.049	G>T	4.361	.113	A
IL6	rs2069840	22651652	7	HapA	.333	C>G	6.624	.036	A
IL6	rs1554606	22651787	7	HapA	.319	G>T	1.085	.581	A
IL6	rs2069845	22653229	7	HapA	.319	A>G	1.085	.581	A
IL6	rs2069849	22654236	7		.024	C>T	NA	NA	NA
IL6	rs2069861	22654734	7	HapA	.056	C>T	4.126	.127	A
IL6	rs35610689	22656903	7		.259	A>G	1.268	.530	A
IL6	HapA1						3.165	.205	
IL6	HapA5						8.421	.015	
IL6	HapA8						2.595	.273	
IL8	rs4073	70417508	4	HapA	.455	T>A	1.750	.417	A
IL8	rs2227306	70418539	4	HapA	.366	C>T	2.107	.349	A
IL8	rs2227543	70419394	4	HapA	.368	C>T	1.239	.538	A
IL8	HapA1						1.750	.417	
IL8	HapA4						1.536	.464	
IL10	rs3024505	177638230	1	HapA	.129	C>T	0.309	.857	A
IL10	rs3024498	177639855	1	HapA	.204	A>G	0.728	.695	A
IL10	rs3024496	177640190	1	HapA	.421	T>C	0.105	.949	A
IL10	rs1878672	177642039	1	HapA	.416	G>C	0.140	.933	A
IL10	rs3024492	177642438	1		.190	T>A	NA	NA	NA
IL10	rs1518111	177642971	1	HapA	.303	G>A	2.862	.239	A
IL10	rs1518110	177643187	1	HapA	.301	G>T	2.973	.226	A
IL10	rs3024491	177643372	1	HapA	.408	G>T	0.021	.989	A
IL10	HapA1						2.703	.259	
IL10	HapA2						3.079	.214	
IL10	HapA8						0.620	.734	
IL13	rs1881457	127184713	5		.210	A>C	2.660	.265	A
IL13	rs1800925	127185113	5		.233	C>T	1.933	.380	A
IL13	rs2069743	127185579	5		.019	A>G	NA	NA	NA
IL13	rs1295686	127188147	5	HapA	.265	G>A	FE	.027	D
IL13	rs20541	127188268	5	HapA	.212	C>T	1.799	.407	A
IL13	HapA1						6.032	.049	
IL13	HapA4						1.479	.477	
IL17A	rs4711998	51881422	6		.346	G>A	1.733	.421	A
IL17A	rs8193036	51881562	6		.327	T>C	0.661	.719	A
IL17A	rs3819024	51881855	6		.372	A>G	0.083	.959	A
IL17A	rs2275913	51882102	6		.361	G>A	2.138	.343	A
IL17A	rs3804513	51884266	6		.023	A>T	NA	NA	NA

# TNF- $\alpha$ Polymorphisms and Anxiety

**Supplemental Table 1** Continued

Gene	SNP	Position	Chromosome	Haplotype <sup>a</sup>	MAF	Alleles	$\chi^2$	P Value	Model
IL17A	rs7747909	51885318	6		.217	G>A	1.817	.403	A
NFKB1	rs3774933	103645369	4		.409	T>C	0.335	.846	A
NFKB1	rs170731	103667933	4	HapA	.358	A>T	0.373	.830	A
NFKB1	rs17032779	103685279	4		.011	T>C	NA	NA	NA
NFKB1	rs230510	103695201	4	HapA	.410	T>A	2.362	.307	A
NFKB1	rs230494	103706005	4	HapA	.434	A>G	0.332	.847	A
NFKB1	rs4648016	103708706	4		.010	C>T	NA	NA	NA
NFKB1	rs4648018	103709236	4		.018	G>C	NA	NA	NA
NFKB1	rs3774956	103727564	4	HapA	.435	C>T	0.231	.891	A
NFKB1	rs10489114	103730426	4		.018	A>G	NA	NA	NA
NFKB1	rs4648068	103737343	4		.363	A>G	0.448	.799	A
NFKB1	rs4648095	103746914	4		.052	T>C	FE	1.000	A
NFKB1	rs4648110	103752867	4		.170	T>A	0.179	.915	A
NFKB1	rs4648135	103755716	4		.061	A>G	FE	.860	A
NFKB1	rs4648141	103755947	4		.180	G>A	1.269	.530	A
NFKB1	rs1609798	103756488	4		.337	C>T	0.532	.766	A
NFKB1	HapA1						2.039	.361	
NFKB1	HapA9						0.409	.815	
NFKB2	rs12772374	104146901	10		.168	A>G	0.011	.995	A
NFKB2	rs7897947	104147701	10		.221	T>G	0.631	.730	A
NFKB2	rs11574849	104149686	10		.070	G>A	2.091	.351	A
NFKB2	rs1056890	104152760	10		.305	C>T	FE	.027	D
TNFA	rs2857602	31533378	6	HapA	.341	T>C	1.473	.479	A
TNFA	rs1800683	31540071	6	HapA	.390	G>A	0.707	.702	A
TNFA	rs2239704	31540141	6	HapA	.335	G>T	1.574	.455	A
TNFA	rs2229094	31540556	6	HapA	.278	T>C	1.163	.559	A
TNFA	rs1041981	31540784	6	HapA	.386	C>A	0.713	.700	A
TNFA	rs1799964	31542308	6	HapA	.224	T>C	FE	.019	R
TNFA	rs1800750	31542963	6		.016	G>A	NA	NA	NA
TNFA	rs1800629	31543031	6	HapA	.149	G>A	2.240	.326	A
TNFA	rs1800610	31543827	6	HapA	.100	C>T	0.083	.959	A
TNFA	rs3093662	31544189	6		.074	A>G	FE	.008	D
TNFA	HapA1						4.818	.090	
TNFA	HapA5						5.194	.074	
TNFA	HapA6						2.166	.339	

Abbreviations: A = additive model; D = dominant model; FE = Fisher's exact test; Hap = haplotype; IFNG = interferon- $\gamma$ ; IL = interleukin; MAF = minor allele frequency; NA = not assayed because SNP violated Hardy-Weinberg expectations ( $P < .001$ ) or because MAF was  $< 0.05$ ; NFKB = nuclear factor  $\kappa\beta$ ; R = recessive model; SNP = single nucleotide polymorphism; TNFA = tumor necrosis factor- $\alpha$ .

<sup>a</sup>The SNPs used to infer the haplotypes for each gene are identified in the "Haplotype" column (eg, for IL13, HapA was inferred using rs1295686 and rs20541).