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A Systematic Review of the Association between Immunogenomic Markers and Cancer-Related Fatigue

LN Saligan¹ and HS Kim¹

¹National Institute of Nursing Research, National Institutes of Health, Bethesda, Maryland, USA

Abstract

Fatigue, which is one of the most commonly reported symptoms in cancer, can negatively impact the functional status and the health-related quality of life of individuals. This paper systematically reviews 34 studies to determine patterns of associations between immunogenomic markers and levels of cancer-related fatigue (CRF). Findings from the longitudinal studies revealed that elevated fatigue symptoms especially of women with early stages of breast cancer were associated with high levels of neutrophil/monocyte, IL-1ra, and IL-6 during radiation therapy; high levels of CD4+, IL-1 β , and IL-6 with stressing stimuli; high levels of IL-1 β during chemotherapy; low NK cell levels after chemotherapy; and presence of homozygous IL-6 and TNF alleles. In the cross-sectional studies, associations between levels of fatigue and immune/inflammatory markers were not consistently found, especially when covariates such as BMI, ethnicity, menopausal status, and educational level were controlled in the statistical analyses. However, a number of genomic markers were observed to be elevated mostly in fatigued breast cancer survivors in the cross-sectional studies. Gaps in knowledge and recommendations for future research are discussed.

Advances in cancer treatment have led to high survival rates and prolonged the natural history of the disease. However, improved survival rates are mitigated by symptoms associated with treatments that lower the health-related quality of life (HRQOL) for survivors. Fatigue is one of the most commonly reported symptoms in cancer with a prevalence rate of 59% to 100% depending on the clinical status of the disease (Weis, 2011).

Cancer-related fatigue (CRF) is defined as a "distressing, persistent subjective sense of tiredness or exhaustion related to cancer or cancer treatment that is not proportional to recent activity and that interferes with usual functioning" (Berger et al., 2010, p. 906). CRF considerably impacts the functional status and HRQOL of individuals by imposing physical limitations and psychological impairments (Curt et al., 2000). Although studies have been conducted to identify associations between immune and inflammatory markers with the severity and intensity of CRF, no causation between a specific biomarker and CRF has been established, contributing to its inadequate clinical management.

The diversity of factors that predispose to CRF suggests that it is a multidimensional symptom that has common related molecular pathways with multiple contributory

Corresponding Author: Leorey N. Saligan, PhD, RN, CRNP, National Institute of Nursing Research, National Institutes of Health, 9000 Rockville Pike, Building 10, Room 2–1339, Bethesda, MD 20892, Phone: 301-451-1685 Fax: 301-480-1413, saliganl@mail.nih.gov.

Conflict of interest

The authors report no conflicts of interest.

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mechanisms (Miaskowski, 2002). The current understanding about the etiology of CRF is based on limited evidence that environmental, genetic, psychological, and physiological factors play important roles in the impairment of oxygen supply, neuromuscular signaling, and the hypothalamus-pituitary-adrenal axis functioning. The goal of this literature review was to systematically review studies that evaluated immunogenomic markers and CRF in order to identify patterns of associations between these variables.

Methods

An initial generic search in PubMed (any date) using the following key words, "Fatigue AND cancer"[title] yielded 867 articles. A basic search query based on the terms from the 867 articles was developed. These key words/phrases include: ("Neoplasms" [Medical Subject Heading (Mesh)]) AND ("Fatigue/blood" [Mesh] OR "Fatigue/cerebrospinal fluid" [Mesh] OR "Fatigue/enzymology" [Mesh] OR "Fatigue/etiology" [Mesh] OR "Fatigue/ genetics" [Mesh] OR "Fatigue/immunology" [Mesh] OR "Fatigue/pathology" [Mesh] OR "Fatigue/physiology" [Mesh] OR "Fatigue/physiopathology" [Mesh] OR "Fatigue/urine" [Mesh]. This basic search query yielded 1992 articles. The search was further refined, limiting the search using specific terms to include, ("Cytokines" [Mesh] OR "Receptors, Cytokine" [Mesh]) OR ("Antigens, Surface" [Mesh] OR "Receptors, Immunologic" [Mesh] OR "Immunologic Factors" [Mesh]) OR "Antigens" [Mesh]) OR "Lymphocytes" [Mesh]) OR "Inflammation" [Mesh]) OR "Interleukins" [Mesh]) OR "Leukocytes" [Mesh]) OR "Biological Markers" [Mesh]) OR ("Gene Expression/genetics" [Mesh] OR "Gene Expression/immunology" [Mesh] OR "Gene Expression/physiology" [Mesh])) OR "Genotype" [Mesh]) OR ("Immunity, Cellular/genetics" [Mesh] OR "Immunity, Cellular/ immunology" [Mesh] OR "Immunity, Cellular/physiology" [Mesh]). This refined search yielded 348 articles. Review, editorial, case studies, and meta-analysis articles were excluded from the list, which reduced the number of indexed articles to 63. To retrieve recent, unindexed articles in PubMed, the following key words were used, Fatigue AND (Cancer OR neoplasms) AND protein* OR biomarker* OR receptor* OR genes OR gene OR genet* OR immunolog* (*truncated words). The search for unindexed articles was limited to articles published in the last 90 days. This search yielded 35 articles. The abstracts of these 98 articles (indexed=63, unindexed=35) were visually reviewed to determine if they met the following inclusion criteria: (a) tested a relationship between specific immune agents and CRF or (b) mentioned the function, concentration, and/or activity of an immune marker and its association with the increasing level, duration, or worsening in intensity of CRF.

Results

The 34 articles that met the eligibility criteria were reviewed. The earliest article appeared in 2001 and 73.5% (n = 25) of the articles were published from 2006 to the present. Eleven (32%) of the studies used longitudinal designs, while 23 (68%) were cross-sectional. Eighteen (53%) studies enrolled only women subjects, most with breast cancer (94%). Of these, 24% studied women in early stages of breast cancer and 65% studied breast cancer survivors (BCS). About 39% of studies (N = 7/18) investigating BCS were written by one research team who screened participants from the same subject data pool. Sixteen (47%) studies enrolled both men and women participants, with eight (50%) investigating participants with various types of cancer and seven (44%) focusing on terminal cases. One study enrolled pediatric patients (Vallance et al., 2010) and one study enrolled only male participants (Orre et al., 2009). Findings on the relationships between immunogenomic markers and fatigue are presented for the longitudinal and cross-sectional studies separately, according to five categories: systemic inflammatory markers, signals of immune response, concentrations of cytokines, markers of cytokine activity, and genomic markers.

Longitudinal studies

There were 10 studies that serially collected biologic samples from study subjects and explored the associations between levels of immunogenomic markers and fatigue at different time points. Seven of these studies showed significant associations between fatigue and an immunogenomic marker. Eleven different fatigue measures were used in the 10 longitudinal studies, with one study using multiple fatigue scales (Panju et al., 2009) and another using an interview to measure fatigue (Olson et al., 2002). Three of these fatigue questionnaires were administered and validated in non-English languages (Ahlberg et al., 2004; Geinitz et al., 2001; Reinertsen et al., 2011). No other questionnaire, except for the Functional Assessment of Cancer Therapy – Fatigue (FACT-F) was used twice, by two longitudinal studies (Panju et al., 2009; Wratten et al., 2004). The FACT-F is the most preferred instrument to measure CRF because it has been used extensively in large studies, has been shown to be sensitive to clinically significant changes in fatigue, and has robust psychometric properties (Minton et al., 2009). No other fatigue measure was used by more than one study.

In four of the ten longitudinal studies, subjects were followed during their radiation therapy (Ahlberg et al., 2004; Bower et al., 2009; Geinitz et al., 2001; Wratten et al., 2004), three studies followed subjects during chemotherapy (Mills et al., 2005; Panju et al., 2009; Vallance et al., 2010), and in one study, subjects were followed while they were receiving chemotherapy with or without radiation as adjuvant treatment (Olson et al., 2002). One study followed BCS pre and post stress testing (Bower et al., 2007), and another followed patients after completing their adjuvant therapy and 2–3 years after that initial assessment (Reinertsen et al., 2011). A number of studies controlled for covariates in data analyses. Body mass index (BMI) was the most commonly controlled covariate in the longitudinal studies, followed by depressed mood and age.

Systematic Inflammatory Markers—Four longitudinal studies evaluated associations of systemic inflammatory markers with levels of fatigue (Bower et al., 2007; Olson et al., 2002; Reinertsen et al., 2011; Wratten et al., 2004) and all used whole blood samples. Two of the longitudinal studies followed patients with early breast cancer stages during their radiation therapy (RT). Of these, one demonstrated a significant association between high levels of C-reactive protein (CRP) and increased fatigue duration (Bower et al., 2009) and the other study showed a significant association between high levels of CRP and fatigue at baseline but not during RT (Wratten et al., 2004). Significant association was also noted between higher fatigue levels and elevated neutrophil count prior to cancer treatment in patients with terminal cases (Olson et al., 2002) and during RT in women with early breast cancer (Wratten et al., 2004). An elevated monocyte count also was significantly correlated with higher fatigue levels at baseline and during RT after controlling for BMI (Wratten et al., 2004). Lymphocyte count did not show an association with fatigue levels (Wratten et al., 2004).

Signal of Immune Response—CD4⁺ lymphocyte was the only signal of immune response measured using a longitudinal design in patients with early stage breast cancer, and it showed a significant association with fatigue level (Bower et al., 2007). These systemic inflammatory markers and the signal of immune response are measures of cumulative activity of pro-inflammatory cytokines such as IL-6 (Bower et al., 2009). The positive associations between these systemic inflammatory markers, signal of immune response, and fatigue may reflect subclinical inflammatory changes related to the disease and/or treatment. Another possible explanation for these associations may be related to chance effect from the multiple analyses brought about by the use of several covariates.

Concentrations of Cytokines—Eleven cytokines were measured in eight longitudinal studies, with IL-6, TNF-α, and IL-1β being the most measured cytokines. These three proinflammatory cytokines showed inconsistent associations with fatigue levels; however, IL-6 levels from serum and blood cell supernatant samples of breast cancer patients with early disease showed positive significant associations with fatigue levels during RT (Wratten et al., 2004) and with induction of stress (Bower et al., 2007). However, plasma levels of IL-6 and fatigue symptoms were negatively associated in women with early stages of uterine cancer before and during their RT (Ahlberg et al., 2004). This significant association with fatigue levels was not observed when serum IL-6 levels were measured from cancer patients with terminal disease (Olson et al., 2002) nor in older individuals (>50 years) with acute myeloid leukemia (Panju et al., 2009), suggesting that the association of IL-6 and fatigue in these longitudinal studies is influenced by the stage and type of cancer and not the specific type of sample used. TNF-α was not associated with fatigue in three longitudinal studies that followed patients during cancer treatment (Ahlberg et al., 2004; Olson et al., 2002; Panju et al., 2009). Higher levels of IL-1β were associated with higher levels of fatigue in women with early stage breast cancer when stress was induced (Bower et al., 2007). However, a significant association between IL-1β and fatigue was not observed in a German study of women with breast cancer receiving RT (Geinitz et al., 2001), which might be related to the inability to find a change in fatigue score during RT using the German-version of the Fatigue Assessment Questionnaire (Geinitz et al., 2001). The negative associations found in the above mentioned studies may be related to the variability in laboratory methodologies used, limited biological half-life of cytokines, or insufficient diffusion of cytokines from peripheral sites into the blood stream (Fuchs et al., 1988).

Marker of Cytokine Activity—IL-1ra was the only marker of cytokine activity measured in the study by Bower and colleagues (2009), where it showed a positive significant association with fatigue levels. Pro-inflammatory cytokines such as IL-6 and IL-1 β are known as key mediators of neuroimmune interactions, and are thought to play a role in the development of fatigue (Bower et al., 2009). These pro-inflammatory cytokines, especially the IL-1 family of cytokines (IL-1 α , IL-1 β) have also been suggested to play an important role in breast cancer progression (Miller et al., 2000; Pantschenko et al, 2003). Animal data strongly suggest that IL-1 is the crucial factor in determining the balance between immunity and inflammation in tumor progression (Voronov et al., 2010). The elevation of IL-1ra in the reviewed articles may be related to its role as a natural inhibitor of IL-1 and is necessary to counter the pro-metastatic activities of IL-1 and IL-6 (Apte et al., 2006).

Genomic Markers—Although there were two longitudinal studies that measured genomic markers, neither found a significant association between a genomic marker and fatigue (Reinertsen et al., 2011; Vallance et al., 2010). DNA extracted from peripheral blood cells of children with acute lymphoblastic leukemia treated with dexamethasone explored polymorphisms from 3 genes (*AHSG, IL6, POLDIP3*), which were found to be related to sleep disturbance but not to fatigue (Vallance et al., 2010). Single nucleotide polymorphisms (SNPs) of inflammation-related genes including, IL1β (rs16944), IL6 (rs1800795), IL6 receptor (rs4129267, rs4845617, rs2228145), and CRP (rs2794521) did not show significant correlations with persistent fatigue in BCS; however, the CRP gene SNP rs3091244 was associated with serum hsCRP level among persistent fatigued BCS (Reinertsen et al., 2011). The study by Vallance and colleagues (2010) study was the first to explore associations between genomic markers and fatigue in pediatric cancer population, therefore more investigation is necessary to validate the results. Reinertsen and colleagues (2011) grouped subjects into persistent fatigued, defined as having fatigue for more than 6 months in duration, much different than the proposed definition CRF (Cella et al., 2001). This study

also collected samples at 2 time points that were two to three years apart, which should be considered in interpreting their results.

Summary—Findings from the longitudinal studies reveal that elevated fatigue symptoms especially of women with early stages of breast cancer were associated with high levels of neutrophil/monocyte, IL-1ra, and IL-6 during RT; as well as high levels of CD4 $^+$, IL-1 β , and IL-6 with stressing stimuli. There were too few longitudinal studies that focused on terminal (Olson et al., 2002) and pediatric cases (Vallance et al., 2010) to identify a trend in association between CRF and immunogenomic markers. In addition, no associations were found between fatigue levels and most genomic markers (Reinertsen et al., 2011; Vallance et al., 2010). Findings from these longitudinal studies suggest that the relationship between fatigue levels and inflammation is complex and not easily discernible.

Cross-sectional studies

Twenty four studies collected biologic samples from subjects at one study time point and explored associations between the levels of immunogenomic markers and fatigue. Twenty two studies showed significant associations between fatigue and an immunogenomic marker. Of these 24 studies, 13 enrolled women with breast cancer and 46% of these 13 studies were conducted by one research team using participants from the same data pool of BCS (Bower et al., 2002; Bower et al., 2003; Bower et al., 2011a; Bower et al., 2011b; Collado-Hidalgo et al., 2006; Collado-Hidalgo et al., 2008). Five studies enrolled patients with terminal disease (Kwak et al., 2012; Inagaki et al., 2008; Minton et al., 2012; Rausch et al., 2010; Scheede-Bergdahl et al., 2012).

Fourteen different fatigue questionnaires were used in the cross-sectional studies and six of these were administered and validated in languages other than English. One non-English fatigue measure was author-developed (Inagaki et al., 2008) and reliability and validity had been established for it in a previous study (Okuyama et al., 2000). The vitality scale of the Medical Outcomes Study Short Form (SF)-36 was the most frequently used to categorize fatigued from non-fatigued subjects (Bower et al., 2002; Bower et al., 2003; Bower et al., 2011a; Collado-Hidalgo et al., 2006; Rausch et al., 2010). Although the SF-36 vitality scale has been found to have good internal consistency (0.85 – 0.87) using large samples, and a high test-retest reliability (0.80) over a two-week period, one concern is that this four-item scale with two items asking about energy and two asking about fatigue might be an inadequate representation of fatigue (O'Connor, 2004). Age was the most commonly controlled covariate in these cross-sectional studies. Other covariates controlled during statistical analyses included gender, body mass index (BMI), type of cancer treatment received, time since completion of cancer treatment, depression, and behavioral status such as smoking, caffeine, and alcohol uses.

Systemic Inflammatory Markers—Of the 24 cross-sectional studies, 13 evaluated associations of systemic inflammatory markers with levels of fatigue. White blood cells (Alexander et al., 2009; Bower et al., 2003; Kwak et al., 2012; Landmark-Høyvik et al., 2009; Orre et al., 2011; Paddison et al., 2009), CRP (Alexander et al., 2009; Booker et al., 2009; Bower et al., 2011b; Kwak et al., 2012; Minton et al., 2012; Orre et al., 2009; Orre et al., 2011; Schroecksnadel, 2007; Scott et al., 2002), and lymphocytes (Alexander et al., 2009; Bower et al., 2003; Collado-Hidalgo et al., 2006; Landmark-Høyvik et al., 2009; Paddison et al., 2009) were the most commonly measured. Results were inconsistent in regard to significant associations between levels of these systemic inflammatory markers and fatigue regardless of type and stage of cancer. WBC levels were significantly elevated in fatigued BCS (Alexander et al., 2009; Landmark-Høyvik et al., 2009; Orre et al., 2011) and in lung cancer patients with stage IIIb-IV disease (Paddison et al., 2009). However, this

significant association was not observed in other studies that enrolled BCS (Bower et al., 2003) nor in a study of patients with various types of terminal cancers (Kwak et al., 2012). Although CRP levels were significantly elevated in patients with high fatigue symptoms in five studies (Alexander et al., 2009; Booker et al., 2009; Kwak et al., 2012; Orre et al., 2009; Schroecksnadel, 2007), two other studies did not find empirical support for this association (Bower et al., 2011b; Minton et al., 2012). Three studies also showed an association between high levels of lymphocytes and fatigue in BCS (Bower et al., 2003; Collado-Hidalgo et al., 2006; Landmark-Høyvik et al., 2009), but two other studies did not show similar significant associations between the two variables (Alexander et al., 2009; Paddison et al., 2009).

A common observation that may explain these inconsistent associations between levels of fatigue and systemic inflammatory markers was the type of covariates used in the analyses. Positive, significant associations between levels of fatigue and the systemic inflammatory markers (WBC, CRP, and lymphocytes) were generally found after controlling for covariates such age, gender, and time since completion of cancer treatment. However, significant associations, especially of fatigue with WBC and CRP, were not observed when BMI was added as a covariate to the statistical analyses (Bower et al., 2003, Bower et al., 2011; Kwak et al., 2012; Minton et al., 2012). Low tryptophan concentrations and high kynurenine/tryptophan ratio showed significant associations with high fatigue levels in patients with malignant disease (Schroecksnadel et al., 2007). Further investigation is necessary to determine the role of tryptophan, a precursor of serotonin in CRF, which can be valuable information for CRF management.

Signals of Immune Response—Four cross-sectional studies explored associations between levels of fatigue and signals of immune response, all in breast cancer survivors (Bower et al., 2002; Bower et al., 2003; Collado-Hidalgo et al., 2006; Von Ah et al., 2008). All these studies showed significant associations between levels of fatigue and signals of immune response. Although NK cell activities were noted to be lower in fatigued BCS (Bower et al., 2002; Von Ah et al., 2008), CD3+ and CD4+ T lymphocytes were higher in the same population in one study (Bower et al., 2003).

Concentrations of Cytokines—Seven cross-sectional studies explored the relationship between levels of fatigue and cytokines (Collado-Hidalgo et al., 2006; Kwak et al., 2012; Inagaki et al., 2008; Orre et al., 2009; Orre et al., 2011; Scheede-Bergdahl et al., 2012; Von Ah et al., 2008). The pro-inflammatory cytokine, IL-6, was the most common cytokine investigated. Higher plasma levels of IL-6 were significantly associated with fatigue symptoms in cancer patients with terminal disease (Inagaki et al., 2008). IL-6 production was also elevated in an *ex-vivo* experiment of stimulated monocytes from fatigued BCS; however, plasma IL-6 levels from the same subjects were not significantly different from non-fatigued BCS samples (Collado-Hidalgo et al., 2006). Four other cross-sectional studies did not find significant associations between levels of fatigue and IL-6 (Kwak et al., 2012; Orre et al., 2009; Orre et al., 2011; Scheede-Bergdahl et al., 2012).

The inconsistency in association between the levels of IL-6 in the blood and fatigue might be related to the fatigue measure used. Although an author-developed questionnaire showed significant association between these two variables, the rest of the cross-sectional studies using more psychometrically sound scales failed to document an association. TNF- α was the other cytokine that was most measured among the cross-sectional studies and its blood level did not show an association with fatigue in three studies (Kwak et al., 2012; Scheede-Bergdahl et al., 2012; Von Ah et al., 2008).

Markers of Cytokine Activity—Four cross-sectional studies investigated the association between levels of fatigue and markers of cytokine activity. Three were written by one

research team who screened participants from the same data pool of BCS (Bower et al., 2002; Bower et al., 2011; Collado-Hidalgo et al., 2006). IL-1ra and sTNF-RII were the two signals of immune response that were most measured in these eight cross-sectional studies. Inconsistent results in the association between levels of fatigue and these two signals of immune response might be related to confounders adjusted during analysis. The positive, significant association between levels of fatigue and IL-1ra persisted even after controlling for age, BMI, depressive symptom scores, time since completion of cancer treatment (Collado-Hidalgo et al., 2006) and behavioral factors such as smoking, caffeine, and alcohol use (Bower et al., 2002). However, these positive, significant associations between fatigue scores with IL-1ra and sTNF-RII were not observed when ethnicity, menopausal status (Collado-Hidalgo et al., 2008), and educational level (Orre et al., 2011) were added as covariates. One study showed a positive association between level of fatigue and sTNF-RII in breast cancer survivors who received chemotherapy as their primary cancer treatment or initial therapy to treat their cancer, but not in those who received other cancer treatments as primary therapy (Bower et al., 2011b).

Genomic Markers—Seven cross-sectional studies exploring associations between levels of fatigue and genomic markers showed positive, significant associations between these markers and fatigue (Aouizerat et al., 2009; Bower et al., 2011a; Collado-Hidalgo et al., 2008; Fernandez-de-las-Penas et al., 2011; Landmark-Høyvik H et al., 2009; Miaskowki et al., 2010; Rausch et al., 2010). TNF and IL-6 alleles extracted from DNA of archived buffy coat samples were significantly associated with fatigue levels. Both studies used the same sample population. Common, homozygous (AA) alleles of IL-6 were associated with higher levels of evening and morning fatigue symptoms among oncology patients and their family caregivers (Miaskowski et al., 2010). Higher morning fatigue, but not evening fatigue was also noted with homozygous (GG) alleles of the TNF-α gene in the same subjects (Aouizerat et al., 2009). SNPs of several cytokines including IL-1β (rs1143633, rs2853550), IL-1RN (rs397211, rs4252041), and IL-10 (rs1878672, rs3021094) showed significant associations with fatigue levels in lung cancer survivors (Rausch et al., 2010). PLOD1 (procollagen-lysine 1, 2-oxoglutarate 5-dioxygenase 1), a gene involved in glycoprotein metabolism, and the NPCDR1 gene (nasopharyngeal carcinoma, downregulated 1) were differentially expressed in fatigued BCS (Landmark-Høyvik H et al., 2009). Overrepresentation of CC alleles in the IL1B-511 (C/T) polymorphism and elevated occurrence of homozygosity for both the variant C allele and the wild type G allele of the IL6 – 174 (G/C) polymorphism were found to be independent predictors of CRF among breast cancer survivors (Collado-Hidalgo et al., 2008). Genes associated with activation of inflammatory cytokines, chemokine signaling, activation of transcriptions and vascular growth factor, as well as NF-κB response were differentially expressed in fatigued BCS (Bower et al., 2011a). Furthermore, genotypic characterization of BCS showed that specific catechol-O-methyltransferase (COMT) genotypes (Valine (Val)/Methionine (Met) and Met/ Met) were significantly correlated with higher fatigue scores compared to survivors with Val/Val genotype (Fernandez-de-las-Penas et al., 2011).

Summary—Findings from the cross-sectional studies demonstrate that elevated fatigue symptoms had positive, significant associations with systemic inflammatory markers such as WBC, CRP, and lymphocytes, however, these significant associations failed to persist after including BMI as a covariate in the analyses (Bower et al., 2003, Bower et al., 2011; Kwak et al., 2012; Minton et al., 2012). Moreover, associations between levels of fatigue and signals for immune response (IL-1ra, sTNF-RII) also failed to persist when ethnicity, menopausal status, and educational level were added as covariates to the statistical analyses (Collado-Hidalgo et al., 2008, Orre et al., 2011), and when these signals were measured in BCS who received primary cancer treatments other than chemotherapy (Bower et al.,

2011b). A number of studies did not find significant associations between levels of fatigue and the pro-inflammatory cytokine, IL-6 (Kwak et al., 2012; Orre et al., 2009; Orre et al., 2011; Scheede-Bergdahl et al., 2012). A marker of IL-6 activity (IL-6R) was found to be elevated in stimulated cells from fatigued BCS (Bower et al., 2002).

All genomic markers explored by the seven cross-sectional studies showed positive, significant associations with fatigue (Aouizerat et al., 2009; Bower et al., 2011a; Collado-Hidalgo et al., 2008; Fernandez-de-las-Penas et al., 2011; Landmark-Høyvik H et al., 2009; Miaskowki et al., 2010; Rausch et al., 2010). Results from cross-sectional immunogenomic studies provide a pattern that subclinical inflammation and immune dysregulation are observed in patients who had a tendency to get fatigued. However, because of the research design used, these findings fell short in identifying whether the experience of fatigue was related to cancer and/or its treatment. Table 1 summarizes the studies demonstrating significant associations between CRF and immunogenomic markers and Table 2 reports on the studies that did not find empirical support for such an association.

Discussion

The goal of this review was to determine patterns of associations between immunogenomic markers and CRF. In the longitudinal studies, there were trends of associations between levels of CRF and markers of inflammation and immune response, especially in women with early stage of breast cancer. It is premature to specify potential biomarkers for CRF based on the findings because all results were based on associations and therefore do not prove causation. However, this review provides empirical support for the association between high levels of CRF and elevated systemic inflammatory markers (CRP, neutrophils, monocytes, lymphocytes); increased signal of immune response (CD4 $^+$); high cytokine (IL-6, IL-1 β) concentrations; and increased markers of cytokine activities (IL-1ra, sTNF-RII).

Elevation of inflammatory/immune markers has been observed with cancer progression and cancer treatment. Inflammation is considered a direct cause of certain types of malignancies (Ruffell et al., 2010) and some tumors attract inflammatory cells to promote its growth and progression (Whiteside, 2006). Tumor cells are infiltrated by immune cells, predominantly by CD4+ T cells, which are recruited by chemokines and cytokines (Whiteside, 2008). These tumor cells have been shown to induce immune cell dysfunction by interfering with signal transduction, cytokine production and proliferation, and cell migration (Whiteside, 2010). On the other hand, cancer treatment has been shown to cause long-lasting imbalance of the immune system, resulting in a chronic inflammatory state (Ma et al., 2010), as seen by the elevated levels of pro-inflammatory cytokines even during survivorship (Bower et al., 2007; Reinertsen et al., 2011). These pro-inflammatory cytokines are produced in large amounts by monocytes (Fieren, 2012), which were found to be elevated in the studies included in this review (Paddison et al., 2009; Wratten et al., 2004). These pro-inflammatory cytokines are known to act on brain structures to alter behavior; and variations in their concentrations, especially IL-6, IL-1β, and TNF-α can lead to sickness behavior including the symptom of fatigue (Fung et al., 2012). Cytokines cross the blood-brain barrier (BBB) using specific transport systems as demonstrated in previous radio-imaging studies (Banks et al., 1995; Konsman et al., 2004). Prostaglandins released in response to cytokine activation can alter BBB permeability (Schenk et al., 2008) leading to increased cytokine movement and leukocyte migration across the BBB (Strey et al., 2002). The pathway involved in the interaction of cytokines and brain structures may help explain pathways behind CRF as displayed in Figure 1.

Gaps in knowledge were also found that limit the ability to draw conclusions related to the associations of immunogenomic markers and CRF. In this section results of the review are

discussed in relation to: (a) gaps in knowledge, (b) genomic findings, (c) study limitations, and (d) recommendations for future research.

Gaps in Knowledge

The first gap identified is the lack of longitudinal studies exploring the associations of immunogenomic markers and fatigue. Only 29% of the studies used a longitudinal design. More longitudinal studies are necessary to prospectively explore the important roles of cancer progression and treatment in the experience of fatigue in this population. Another gap is the lack of a case-definition for CRF. Only 12 of the 34 studies conceptually defined fatigue. In three studies, CRF was defined as a multidimensional concept where increased levels of CRF were associated with elevated levels of cytokines (IL-6, IL-1\(\beta\)) (Inagaki et al., 2008; Panju et al., 2009; Von Ah et al., 2008). A similar association was observed in one study that used the ICD-10 criteria to define CRF as the presence of significant fatigue nearly every day for two weeks in the past month (Alexander et al., 2009). Three studies did not observe a similar association between CRF and cytokine levels, but higher levels of CRF were associated with higher concentrations of systemic inflammatory markers (CRP, neutrophils) when CRF was conceptualized as having both physical and attentional/mental components (Booker et al., 2009; Olson et al., 2002: Scott et al., 2002) or when it was defined as a sense of tiredness that lasted more than 6 months (Orre et al., 2009; Reinertsen et al., 2011). A non-significant or inverse relationship was observed between CRF and levels of cytokine (IL-6) or systemic inflammatory marker (CRP) when CRF was defined as persistent tiredness (Ahlberg et al., 2004; Kwak et al., 2012) or as part of a symptom cluster (Minton et al., 2012). Variations in the associations between CRF and inflammatory/ immune markers are related to the differences in scope of the concept of fatigue being measured and the duration of symptom experience.

A third gap identified is the lack of standard CRF measures that can predict clinical significance. Only 2 of the 11 questionnaires used in the ten longitudinal studies namely, the Fatigue Symptom Inventory (FSI) and the Multidimensional Fatigue Symptom Inventory (MFSI), were reported to be sensitive to changes of disease progression or treatment (Whitehead, 2009). Thus, the other CRF measures used in the longitudinal studies had not been validated for longitudinal designs to be sensitive to cancer progression or effect of cancer therapy. Another important gap that emerged from this review relates to identifying whether fatigue symptoms experienced by cancer patients were related to disease progression, cancer treatments, or other covariates. Inconsistent or non-significant associations were found in studies that enrolled subjects with terminal cases or older individuals or when BMI, ethnicity, and menopausal status were included as covariates in the analyses. A final gap of knowledge is the lack of consistent associations between concentrations of pro-inflammatory cytokines (IL-6 and TNF-α) and levels of fatigue. These latter associations require further investigation using more sophisticated approaches.

After addressing the aforementioned knowledge gaps, the use of genomic and proteomic technologies in identifying the roles of genes, proteins, and environment in CRF may best describe pathways that might play a crucial role in the development of fatigue in cancer populations. These approaches have been proven successful in identifying mechanisms in other symptoms such as depression (Keers R, Uher R, 2011) and pain (Kaszas et al., 2012).

Genomic Findings

Genomic technology is a novel approach for providing information about possible pathways that may explain development of CRF and a closer attention to the findings reported by the reviewed articles is warranted. The findings related to the genetic association studies conducted by Collado-Hidalgo et al. (2008) and Bower et al. (2011a) must be considered

with caution because of two critical limitations: both used small samples and both further stratified their samples during analysis. For example, the genetic association study investigated 33 fatigue patients and 14 non-fatigue patients and further subdivided the sample into two ethnic groups (whites and non-whites) to control for ethnicity. However, a closer investigation of the non-white subjects especially those who were fatigued, revealed that 40% were Asians. Furthermore, there were no Asian subjects in the non-fatigued group (Collado-Hidalgo et al., 2008), which made it more difficult to interpret the results.

Aouizerat et al. (2009) and Miaskowski et al. (2010) reported a genetic association between certain SNPs and fatigue. Both used the same population of cancer patients and caregivers. Even though these studies found significant associations, there were some limitations in their similar study design. The authors combined cancer patients and their family members for the analysis, which could have led to different types of fatigue being analyzed: cancer-related fatigue and fatigue not related to cancer. The population stratification is another limitation of these studies considering that categorizing different races into a single, non-white, ethnic category would provide additional complexity to genomic analysis.

The Norwegian research team reported significant associations between genomic markers and fatigue in breast cancer survivors (Landmark-Høyvik H et al., 2009; Reinertsen et al., 2011). Among the eight articles reporting on genomic marker association with fatigue, the study conducted by Landmark-Høyvik, H. et al. (2009) deserves a closer inspection because of its large sample size. An initial sample of 403 subjects was reduced to a subset of 137 with chronic fatigue assessed at two different time points. These 137 subjects were further stratified into fatigue and non-fatigue groups to control for differences in anxiety and depression levels. This study showed an association between a single gene expression using blood and fatigue levels using linear analysis. However, more sophisticated gene set enrichment analyses (GSEA) using pathways from the Molecular Signature Database (MSigDB) revealed a difference of gene expression in inflammatory process and immune system. The study conducted by Reinertsen et al. (2011) showed a weak association between a C-reactive protein encoding gene SNP (rs3091244) and fatigue. Because the uncorrected p-value of the reported association was only 0.02, it may be considered as a false positive result.

Findings from a study by Fernandez et al. (2011) also warrant further consideration. This study reported a positive association between high fatigue levels in breast cancer survivors with a Met/Met COMT 158 when these subjects also complained of more pain in the neck and shoulder areas. Considering high minor allele frequency of COMT 158, the sample size of this study (N = 128) seems to be appropriate. This COMT SNP has been studied extensively including its role in experimental and clinical symptoms such as opioid responses in cancer patients, but some inconsistencies have been reported (Kambur et al., 2010; Laugsand et al., 2011). The role of COMT SNP in CRF needs to be investigated further.

Recommendations for future research

More longitudinal studies that address the identified gaps are needed to fully advance the investigation of mechanisms related to CRF and to capture the dynamic changes that occur in these immunogenomic markers during cancer progression and treatment. Careful sample and study design selection, utilization of valid and reliable CRF measures that are sensitive to changes in fatigue overtime, and inclusion of relevant covariates in statistical analyses are important considerations in designing future studies. Establishing causation between biomarkers and CRF will not be fully realized without CRF being case defined, clearly measured, and clinically translated. Identifying a biomarker of CRF will be immensely

beneficial not only in the clinical management of CRF but also in improving treatment outcomes of individuals with cancer.

Conclusion

This review identified some patterns of associations between specific immunogenomic markers and fatigue in survivors with early stages of cancer. Inconsistent associations between fatigue and immunogenomic markers were found in subjects with terminal cases of cancer and when other covariates where considered in the analysis. The most important findings of this review are the identification of the gaps of knowledge that must be addressed in order to advance the science of CRF research. Future efforts should focus on defining CRF, reassessing clinical significance of CRF measures especially using longitudinal approaches, and using biomarkers in predicting changes in CRF. Methodologically improved designs can pave the way in understanding etiologic mechanisms and therapeutic targets of CRF.

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Highlights

This review identified patterns of associations between immunogenomic markers and CRF, and gaps in knowledge needed to advance the science of CRF research.

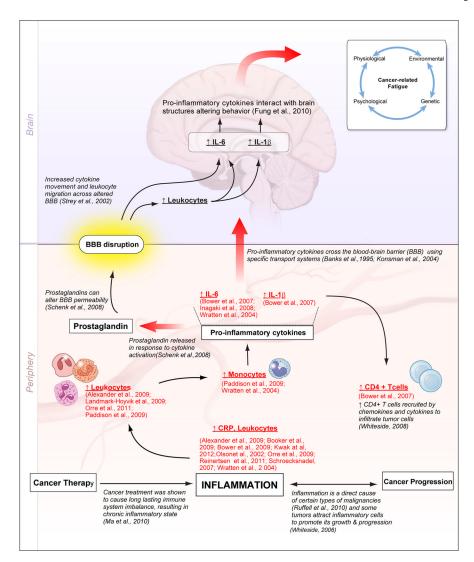


Figure 1. Association of Inflammatory Markers and Cancer-Related Fatigue

The link between inflammatory markers and cancer-related fatigue may be related to the inflammatory state generated by cancer progression and/or cancer therapy. Both conditions trigger an increase in pro-inflammatory cytokine production by white blood cells (especially monocytes). The systemic experience of CRF may be related to the interactions of pro-inflammatory cytokines and immune cells with brain structures that migrate through a disrupted blood-brain barrier altered by pro-inflammatory cytokine-related activities. CRF intensity is dependent on physiological, psychological, genetic, and environmental factors.

Table 1

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gnificant Associations Between Immunogenomic Markers And CRF

	Direction of Association	Positive, significant correlation between IL-6 and fatigue from baseline to 3 weeks into RT	Positive, significant correlation between IL-6 and fatigue from baseline to end of RT	Higher level and greater increase in number during stress test in fatigued group	Increase in lipopolysaccharide (LPS)-stimulated production during stress test in fatigued group	Increased from baseline to recovery from stress test in fatigued group.	Higher with increased fatigue duration	Higher with increased fatigue severity	Higher levels predicted high fatigue in the 4th cycle of chemo, but no association with fatigue when depression was added.	Higher levels associated with high fatigue prechemo in colon cancer patients
	Covariates			Time between blood draw, age, marital status, cancer treatment, body mass index (BMI), and depressed mood score			Sleep disturbance, depressive symptoms,	age, body mass index, hormone therapy	Depression, stage of disease	None mentioned
	Association	r= -0.65	r= -0.54	F(2,46) = 4.0	F(1,20) = 9.3	F(1,20) = 6.1	β = 0.32, SE = 0.14	$\beta = 0.63$, SE = 0.26	beta = 0.468	I = 0.737
	p value	0.006	0.04	0.024	0.006	0.02	0.022	0.016	< 0.03	0.01
	Marker I	Interleukin (IL)-6	<u> </u> -	CD4+	IL-6	π-1β	C-reactive protein (CRP)	L-1ra	Vascular Growth factor (VEGF)	Absolute neutrophil
Longitudinal Studies	Data Collection	Plasma, pre RT, post 30 Gy (+3 wks), post RT (46 Gy, or +5-6	wks), I week post RT	Lymphocytes extracted at baseline, post stress, after 30 minutes of recovery.	Blood supernatant at baseline, post stress, after 30 minutes of	recovery.	Serum drawn at baseline, 4 time	points during K1, 2 time points post RT	Plasma before chemo cycles 1 and 4; 2.5 months between plasma collection	Serum before treatment, midtreatment, 3 and 6 months post treatment
Longitu	Measure	Multidimensional Fatigue Inventory (MFI) (Swedish)		Short Form (SF)36– vitality scale			Fatigue Symptom Inventory (FSI)		Multidimensional FSI (MFSI)	Interview
	Intervention	Radiation Therapy (RT)		Survivors			RT		Chemo (adjuvant or neoadjuvant)	Chemo and/or RT (for rectal and SCLC patients)
	Control	None mentioned		n = 15 (non- fatigued)			Baseline values for each patient		none	none
	Exclusion	Dementia, history of psychiatric disorder		Cancer recurrence, diagnosis with other cancer, history of immunologic or hormonal disease, current medical	ilness, heavy alcohol use		Tobacco use, recurrent cancer,	previous/planned chemo, immunosuppressant use, active illness/ infection	Undergoing bone marrow transplants, metastatic breast cancer, confounding illness (e.g. renal failure, pre-existing anemia)	None mentioned
	Sample	sc B	rain Behav l	m_{mn}^{mn} . Author m	anuscript; avai	lable in P	Q = 28 $Q breast); n$	(20 20 August 01 1/2)	. n = 29 (Same subjects used by Aouizerat et al., 2009)	<i>n</i> = 18
	ncer	erine, 86% stage I, st hysterectomy		east, stages 0, I, or II			sast (stage 0, I, II); state (T1-Tumor 3,	de U, Mets U)	east, stages 1–IIIA	lon, rectal, Small I lung cancer L.C.), NSCLC ages IIIb-IV)

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	Direction of Association	Higher level in patients with non-depressed, chronic fatigue at T1 and subset analysis (higher with persistent fatigue after controlling for covariates)	Fatigue correlated with higher neutrophil level at baseline & week 5 of RT	Fatigue correlated with higher monocyte level at baseline & week 5 of RT	Fatigue correlated with higher CRP at baseline	Decreased in non- fatigued subjects at week 5 of RT	Fatigue correlated with higher IL6 at baseline and week 5 of RT	Fatigue correlated with higher ICAM-1 at baseline		Direction of Association	Higher in patients with morning fatigue	Higher in fatigued group	Higher in fatigued group	Higher in fatigued group
	Covariates	Treament strategies; subset of subjects: BMI, treatment-area related fibrosis	BMI							Covariates	age, TNFA genotype	None mentioned		
	Association	OR = 1.11, 95% CI (1.01 - 1.21)	Baseline (r=- 0.315), week 5 (r=-0.381)	baseline (r= -0.289), wk5 (r=-0.394)	r=-0.456	Not given	baseline (t = -0.322), wk5(t = -0.367)	r= -0.311		Association	t = -2.22	Not given	Not given	Not given
	p value	0.03	Baseline (p=0.03,), week 5 (p=0.01)	baseline (p=0.05), week 5 (p=0.01)	p<0.01	0.04	baseline (p=0.05), wk5(p=0.03)	p=0.04		p value	0.02	0.021	0.015	0.04
	Marker	hsCRP	Neutrophil	Monocyte	CRP	Fibroblast growth factor eta	IL6	Intercellular adhesion molecule -1 (ICAM-1)		Marker	Homozygous allele of TNFA genotype (i.e., GG)	White blood cell (WBC)	CRP	Basophil
Longitudinal Studies	Data Collection	Deoxyribonucleic acid (DNA) from peripheral blood, blood drawn 2-3 years apart for 2 timepoints	Blood before RT, weekly during RT, then 2 & 6	weeks post K.I.		Serum before RT, weekly during RT, then 2	& 0 weeks post RT		Cross-sectional Studies	Design	Archived buffy coat DNA	Peripheral Blood		
Longit	Measure	Fatigue Question naire (FQ - Norwegian)	Functional Assessment of Cancer Therapy –Fatigue (FACT-F)						Cross-se	Measure	Lee Fatigue Scale (LFS)	Brief Fatigue Scale		
	Intervention	None	Adjuvant RT post surgery							Intervention	RT	Survivors, 3 months -2	year post primary therapy	
	Control	None fatigue d, n =120,	None mentioned							Control	n = 103 (family caregivers)	n = 104 (non-	iaugueu)	
	Exclusion	Breast cancer recurrence, other cancer (except melanoma, ovarian cancer in situ), depression in TI		surgery, mstory or breast irradiation treatment, metastatic disease						Exclusion	Metastatic disease, more than one cancer diagnosed, diagnosed sleep disorder	Pregnancy, other	disease, confusion,	dementia
	Sample	Timepoint (T1)=302, Timepoint 2 (T2)=236 (175) eliminated for for for matigue only min I time agoint). Matigued, Matigued, Matigued, Matigued, Matigued,	25 	thor manu	ıscript;	available	in PMC 2	013 Augu	st 01	Sample	<i>n</i> = 185	n = 60	(Taugue)	
	ncer	east, stages II and III	east, early stage							ncer	rious, non-metastatic	east, stages 1–IIb		

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					Longitue	Longitudinal Studies					
ncer	Sample	Exclusion	Control	Intervention	Measure	Data Collection	Marker	p value	Association	Covariates	Direction of Association
ıltiple myeloma, ges I, II, III	n = 56	Other plasma cell dyscrasia	None	None	FACT-F	Peripheral blood	CRP	0.034	$\beta = -0.350$	None mentioned	Significant predictor of fatigue using FACT-F
					Quality of Life Questionnaire (QLQ)-C30			0.003	$\beta = 0.514$		Significant predictor of fatigue using QLQ-C30
east, recruited from a ol of 332 survivors	n = 20 (fatigued)	Change in energy level between 2	n = 20 (nonfatigued)	Survivors recruited from 1994–1997; 5	RAND36, FSI	Serum taken between 8–10	L-1ra	0.006	95% CI = 2% to 89%	caffeine, alcohol use, smoking	46% more in fatigued group
o met englomty eria used by Bower d., 2006	Brain	assessments, cancer recurrence, other cancer, comorbid		years from diagnosis		am, tasung/no alcohol/caffeine/ smoking 12	Soluble tumor necrosis factor (TNF)-RII	0.005	95% CI = 6% to 59%		18% more in fatigued group
	Behav I	medical problem, immune disease, on immunosuppressant,				hours before draw	Neopterin	0.018	95% CI = 1% to 34%		33% more in fatigued group
	Immun.	psychiatric hospitalization in past 6 months,					Natural killer (NK) cells	< 0.05	A(1,33) = 4.33	Smoking	Lower percentage in fatigued group
	Autho	heavy alcohol use					CD45RO:CD45DA	0.05	R(1,33) = 4.01		Higher ratio in fatigued group
m a rs	$ \mathbf{g}_{n} = 19 $ $ \mathbf{g}_{n} = 19 $	Recurrence of breast cancer, history of	n = 18 (nonfatigued)	Survivors recruited from 1994–1997	RAND SF36 fatigue scale grouped subjects to fatigued	Lymphocytes from fasting	Lymphocytes	0.011	95% CI = 7% to 49%,	age, income, ethnicity, BMI,	28% more in fatigued group
o met englomty eria used by Bower d., 2002	script; a	minume disease			and non-tangued during z assessments	00010	CD3+	0.015	95% CI = 6% to 56%	depressed mood, cancer treatment	31% more in fatigued group
	vailabl						CD4+	0.003	95% CI = 15% to 68%		41% more in fatigued group
	e in PN						CD3+/CD56+	0.027	95% CI = 4% to 99%		52% more in fatigued group
II); ne	$ \int_{\mathbf{Q}} \mathbf{d} = 11 $ $ \mathbf{Q} \text{fatigue d),} $	Cancer recurrence, immune disease,	n = 10 (non-fatigued), >70	Survivors, 1–5 years post diagnosis, 21–65	SF-36 vitality scale	Peripheral blood mononuclear	IL 1A, IL 1B, IL 6, OSM, GZMH	<0.001	Not given	Age, time since diagnosis and cancer	Upregulated in fatigued group (>30% difference)
of of subjects ntified by Collado- dalgo et al., 2006.	With SF-36	use use	n Sr-50 laugue scale for 2–3 assessments	years of age		cens	CXCL2, CXCR5, CCL20, CMKLR1	<0.001	Not given	gene analyses	Upregulated in fatigued group (>30% difference)
	Scale for 2-19 assessments						IER3, ZNF331, NR4A2, NR4A3	<0.001	Not given		Upregulated in fatigued group (>30% difference)
							VEGFA, TRGC2, TIGIT, CX3CR1	<0.001	Not given		Upregulated in fatigued group (>30% difference)
							NLRC4, HP, CROP, MGAM	<0.001	Not given		Upregulated in fatigued group (>30% difference)
							NF-kB response elements in promoters of upregulated genes	<0.0001	2.28-fold difference ±0.09		Promoter upregulated in fatigued group (across nine combinations of promoter length)

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	Direction of Association	Promoter under represented in fatigued group (across nine combinations of promoter length)	Upregulated in fatigued group	Higher levels associated with higher fatigue scores with subjects treated with chemotherapy only.	Higher level in fatigued group	Higher level in fatigued group	Increased ex-vivo monocyte production in fatigued group after LPS exposure	Increased ex-vivo monocyte production in fatigued group after LPS exposure	Lower levels on CD14+ cells in fatigued group after exposure of monocyte to toll-like receptor (TLR) 4 ligand LPS	Increased % as fraction of total leukocyte in fatigued group with selective increase in frequency of CD4 T lymphocytes.	Decreased frequency in circulating dendritic cells from fatigued group
	Covariates			Exposure to RT, age, time since treatment completion, BMI	Age, BMI, time since treatment, treatment	mode, depressive symptom scores					
	Association	0.45-fold difference ±0.07	2.72 fold increase	Not given	t(48) = -1.53	t(44) = -4.07	t(45) = -1.813 t = -1.983 $t(29) = 2.195$				t(29) = 2.047
	p value	<0.007	0.041	0.036	0.05	< 0.001	0.049	0.03	0.03	Not given	0.04
	Marker	Glucocorticoid response elements in promoters of upregulated genes	Proinflammatory transcription factor- binding motifs in the promoters of upregulated genes	sTNF-RII	L-1ra	sIL-6R	L-6 (ex vivo)	ΤΝΕ-α (<i>ex vivo</i>)	L-6R (in vivo)	Lymphocytes	Myeloid dendritic cells (HLA-DR+/CD11c+/CD14 dim)
Longitudinal Studies	Data Collection			Plasma	Plasma		Monocytes			Peripheral blood mononuclear cells (PBMCs)	
Longit	Measure			FSI	SF36-vitality scale grouped subjects to fatigued and non-	ratigued during z–5 assessments					
	Intervention			Completed primary treatment within 3 months and not started on endocrine treatmen	Survivors, 1–5 years post diagnosis						
	Control			None mentioned	n = 18 (non- fatigued), score	>/0 in SF-30 fatigue scale					
	Exclusion			Neurologic or immune disease, smokers	<u> </u>		alcohol use, SF-36 fatigue score 50–70, advanced cancer stage, >5 years post treatment				
	Sample		Brain I	801 = 103 <i>efiav Immun</i> . Auth		SK 50 score Swith SF-36 dratigue	ಲ್ಲ ಇvailable in	PMC 2013	August 01.		
	ncer			ast, stages 0-IIIA, ected from same fort as Bower et al., Bower et al., 2003; wer et al., 2011a; llado-Hidalgo et al., 56	east (stages 0,I, II); ruited from same	Bower et al., 2002 Bower et al., 2003 Bower et al., 2003					

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					Longin	Longitudinal Studies					
ncer	Sample	Exclusion	Control	Intervention	Measure	Data Collection	Marker	p value	Association	Covariates	Direction of Association
							Activated T lymphocytes (CD3+/ CD69+)	0.04	t(29) = 1.077		Decreased frequency in circulating activated T lymphocytes from fatigued group
	Bi		Healthy blood				IL-6R (in vitro)	<0.01 for IL.6 and IL.1β; 0.23 for TNFa; 0.0006 for IL6, IL.1β, TNFa	t(4) = 4.12 for IL6; t(4) = 4.51 for IL1β; t(4) = 1.36 for TNFα; t(4) = 5.18 for all 3 cytokines		Lower IL6R from cell surface of PBMCs of fatigued group after exposure to 3 cytokines
	un = 33 Gatigue d),		n = 14 (non- fatigued), SF36	Survivors, 1–5 years post diagnosis	MFSI	Leukocytes from peripheral blood	IL1B-511– CC alleles	0.008	None given	Age, ethnicity, menopausal status,	Overrepresentation in fatigued group.
	Aatigue Acore <55	ımmunosuppressant	ratigue score				IL 1B-511 – TT alleles	0.008	None given	BMI, depressive symptoms, cancer treatment	Underrepresentation in fatigued group.
llado-Hidalgo et al.,)6	<i>mun</i> . A						IL1B-511 – cytosine	0.052	95% CI = 0.91 to 16.6		Higher prevalence in fatigue group
	uthor manuscript						IL 6 – 174 (G/C)	0.024	95% CI = 1.12 to 17.9		Elevated homozygosity in variant C and wildtype G alleles noted in fatigued group except when covariants were controlled.
	, availal					Plasma	sIL-6R	0.028	None given		Higher levels in fatigued group
east, stage I–IIIA	30 (34 (34 (34 (34 (34 (34 (34 (34 (34 (34	Active cancer, receiving chemo/ RT, breast surgery for cosmetic purpose, inflammatory disease, recurrent cancer, fibromyalgia	Val/Val versus Val/Met versus Met/Met genotypes	Survivors treated with RT and chemo (from 6/2009 to 3/2011), 36–65 years old	Piper Fatigue Scale (PFS-Spanish	Genomic DNA extracted from saliva cell sediments	COMT genotypes Val/ Met and Met/Met:	< 0.01	Not given	Pain intensity	Significantly correlated with higher fatigue scores than those with Val/Val genotype.
rious, terminal cases ruited from '1997–11/1999	7 = 27 (faitigued) (faitigued)	Receiving curative cancer treatment, too ill to answer, cognitive impairment, non-Japanese speakers, only included those that died 6 months after 1st assessment, taking NSAIDS and steroids	n = 19 (nonfatigued)	None	Cancer Fatigue Scale (CFS - Japanese – author developed)	Plasma	IL-6	0.01	β = 0.38	Gender, weight, survival time	Higher levels in fatigued group (correlated w/ physical subscale score, but not with total, affective & cognitive scores).
rious, terminal cases ruited from 6/2009– .010; survival time is s than 6 months	n = 90 (mild= 23, modera te =	Cognitive impairment, chemo/ RT to treat active cancer, hematologic	Mild versus moderate versus severe	None	Brief Fatigue Inventory (BFI -Korean)	Peripheral blood collected within 24 hours from	CRP	0.005	r= 0.29	Age, gender, BMI, cancer site, previous cancer treatment, comorbid disease,	Higher levels in more severe categories of fatigue

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	Direction of Association		Higher in fatigued, non- depressed group	Higher in fatigued, non- depressed group	Higher in fatigued, non- depressed group	Downregulated in fatigued group, regardless of depression status	Upregulated in fatigued, non-depressed group	Higher levels associated with higher fatigue scores	Higher levels associated with higher fatigue scores	Higher levels associated with higher fatigue scores	Associated with higher mean evening fatigue scores	Higher in fatigued group, but lost significance when adjusted for BMI	Higher in fatigued group but lost significance when adjusted for behavior (smoking)
	Covariates	pain score, sleep disorder, dyspnea	None mentioned					None given			Age, gender, IL-6 genotype	Age, BMI, smoking, anxiety, depression, neuroticism	
	Association		Not given	Not given	Not given	Not given	Not given	I = 0.62	r = 0.52	r = 0.41	3.8%	I = 0.18	r = 0.16
	p value		0.0016	0.0059	0.0046	False Discovery Rate (FDR) < 0.20	FDR < 0.20	Not given	Not given	Not given	0.001	< 0.01	< 0.05
	Marker		Leukocytes	Neutrophils	Lymphocytes	PLOD1	NPCDR1	П6	IL-IRA	TNF-a	IL-6 AA genotype	IL-1ra	CRP
Longitudinal Studies	Data Collection	enrolment between 9–11 am	Peripheral blood drawn in 2004			Ribonucleic acid (RNA) from blood collected using PAX tube		Serum drawn at baseline, fatigue measure at	baseline and 1 month of treatment		Archived buffy coat DNA	Plasma drawn from 0800-12 noon; subjects	anowed ngnt breakfast
Longitu	Measure		FQ (Norwegian), responses categorized into high and low	iangue				BFI			LFS	FQ (Norwegian)	
	Intervention		Survivors, treated from 1998–2002 with	adjuvant K1				Chemo			RT	Survivors, recruited from 1980–1998 database, 18–75 years	oid
	Control	categories of fatigue	Low fatigue					Normal control (no N given)			n = 103 (family caregivers)	n = 191 (without chronic fatigue)	
	Exclusion	malignancies, use of psychostimulant, h/ o psychiatric disease, fever, use of antibiotics and antiepileptics, high TSH	Cancer recurrence, >75 years old, other	cancer except basal cell cancer or cancer in situ, prior surgery	tor contralateral breast cancer stage I with no adjuvant	therapy		None mentioned			Metastatic disease, more than one cancer diagnosed, diagnosed sleep disorder	Mental retardation, extragonadal germ cell malignancy	except skin), removal of non- affected testicle related to benign condition, concurrent infection or inflammation
	Sample	30, severe =37)	n = 403	nin Beha	av Imm	oun. Author 1	nanusc	$ \frac{di}{dt} = 54 $ $ \frac{d}{dt} \text{ (baseline)}, $ $ \frac{d}{dt} = 26 \text{ (1)} $	on all the state of the state o	1C 2013 A	\$2 	n = 92 (chronic fatigue)	
	ncer		east (stage II, III, classifiable)					ute myelogenous kemia or myelodys stic syndrome			rious, non-metastatic	sticular	

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				Longitu	Longitudinal Studies					
Exclusion		Control	Intervention	Measure	Data Collection	Marker	p value	Association	Covariates	Direction of Association
Cancer recurrence, >75 years old, other cancer except basal cell cancer or cancer in situ, prior surgery for contralateral	ence, , other basal cancer surgery ral	None	Survivors, treated from 1998–2002 with adjuvant RT; subsample of Reinersten et al., 2011 & Landmark-Høyvik H	FQ (Norwegian)	Blood drawn from 0900-12 noon; subjects allowed light breakfast	Leukocytes	<0.018	Beta = 0.014	Age, educational level	Higher levels associated with higher fatigue scores but lost significance when adjusted for covariates (<i>p</i> = 0.78)
breast cancer stage I with no adjuvant therapy	stage I		et al., 2009		Serum	High sensitivity CRP	< 0.001	Beta = 0.26		Associated with total failgue even after controlling for age and education $(p = 0.02)$
None mentioned	pauc	None	None	Functional Assessment of Chronic Illness Therapy-	Retrospective extraction of	WBC	0.01	$R^2 = 0.27, \beta = 0.41$	Age, gender, time since treatment	Increased with higher fatigue
				Faugue (FACII-F) – 2 items, 1 item from Hamilton Depression scale	clinical results (peripheral blood)	Neutrophil	0.01	$R^2 = 0.28, \beta = 0.43$	completion, hemoglobin	Increased with higher fatigue
						Monocyte	0.05	$R^2 = 0.20, \beta = 0.31,$		Increased with higher fatigue
None mentioned	ioned	None mentioned	None	Lung Cancer Symptom Scale, (LCSS)	DNA from peripheral blood	L-1B rs1143633	Not given	OR estimate = 1.00 – 1.02	Age at diagnosis, sex, smoking status,	Increased with higher fatigue
						L-1B rs2853550	Not given	OR estimate = 1.01 – 1.06	disease stage, treatment modality	Increased with higher fatigue
						IL-1RN rs397211	Not given	OR estimate = $0.97 - 1.00$		Increased with higher fatigue
				SF-8		IL-10 rs3021094	Not given	OR estimate = 1.02 – 1.18		Increased with higher fatigue
						IL-10 rs1878672	Not given	OR estimate = 0.91 – 0.94		Increased with higher fatigue
						IL-1RN rs4252041	Not given	OR estimate = 0.80 – 0.97		Increased with higher fatigue
None mentioned	ntioned	None mentioned	None	Fatigue scale	Serum	Neopterin	<0.01	$I_{S} = 0.274$		Increased with high fatigue
						Tryptophan	<0.05	$I_{\rm S} = -0.179$		Lower concentration associated with increase fatigue
						kyn/trp	<0.01	$I_{\rm S} = 0.276$		High ratio with high fatigue
						CRP	<0.001	$I_{\rm S} = 0.375$		Increased with high fatigue
						Erythrocyte sedimentation rate (ESR)-1	<0.01	$I_{\rm S} = 0.234$		Increased with high fatigue

	Sa	ligan an	d Kim		
	Direction of Association	Increased with high fatigue	High levels associated with high fatigue scores	Higher levels associated with high fatigue scores	Lower levels associated with high fatigue scores
	Covariates		None mentioned	Type of adjuvant therapy, mood, network support, satisfaction, cortisol, perceived stress, optimism	
	Association	$I_{\rm S} = 0.241$	Not given	r = 0.32	r 0.20
	p value	<0.01	0.011	0.01	0.01
	Marker	ESR-2	CRP	IL-1β	NK cell activity
Longitudinal Studies	Data Collection		Peripheral blood collected from 1/1995–11/1998	Blood collected 9–33 days post surgery (lumpectomy or mastectomy) before adjuvant therapy	Mononuclear cells collected 9-33 days post surgery (lumpectomy or mastectomy) before adjuvant therapy
Longitu	Measure		QLQ-C30	Revised Piper Fatigue Scale (rPFS)	
	Intervention		None	Adjuvant chemo + RT	
	Control		None	None mentioned	
	Exclusion		Infection	Psychiatric disorder, dementia, history or current substance abuse, thyroid issues, immune disorders, immunosuppression, immunosuppression,	oner cancer except noninvasive type
	Sample	progressing = 86)	n = 106	= 44	nmun. Author manuscrip
	ncer		perable NSCLC, ges III and IV	east, stage O–IIIa	

trp = kynurenine to trygishan ratio typishan ratio

Table 2

Studies With Non-Significant Associations Between Immunogenomic Markers And CRF

	Covariates	None mentioned		None mentioned				None mentioned		None mentioned										Treatment strategies; subset of subjects: BMI, treatment-area related fibrosis
	Association	t = 0.20 (baseline - 30 Gy), $t = -0.19$ (baseline to 46 Gy)	Low concentration, cannot estimate	No association with fatigue				r = 0.19 - 0.61	Not given	No association with fatigue level and	seventy				r=0.33 (FACT-F), r=0.34(QLQ-C30)	r=0.332	r=0.33	Strongest trends of change across	cnanges in tangue severity	No association with fatigue
Longitudinal Studies	p value	0.51 (baseline - 30 Gray), 0.50, (baseline to 46 Gray)	Not given	Not given				0.12-0.62	Not given	Not given	Not given	Not given	Not given	Not given	0.08 (FACT- F), 0.07 (QLQ-C30)	650.0	60.0	Not given	Not given	0.59-0.92, subset = 0.57- 0.7, T2= 0.6- 0.64
	Marker	Tumor necrosis factor (TNF)-α	Interleukin (IL)-1	IL-1β	IL-6	TNF-a.		IL-6	TNF-α	Interferon (IFN)- γ	IL-2	II8	TNF-a	Monocyte chemotactic protein-1	П5	IL-6	IL-10	Monokine induced by IFN- γ	IL-4	IL6R mRNA
Longitudinal Studies	Data Collection	Plasma, pre RT, post 30 Gy (+3 wks), post RT (46 Gy, or +5-6 wks), 1 week post RT		Serum pre RT, end of	weeks 1–5 during K1, 2 months post RT			Serum before treatment,	midtreatment, 3 and 6 months post treatment	Serum T1 (pre	treatment, between treatment, upon	completion of full treatment); T2 (4–6	weeks post T1)							Deoxyribonucleic acid (DNA) from peripheral blood, blood drawn 2–3 years apart for 2 timepoints
	Measure	Multidimensional Fatigue Inventory (MFI) (Swedish)		Fatigue Assessment	(FAQ), German	version		Interview		Functional	Therapy –Fatigue	(FACI-F) and Edmonton Symptom	Assessment System (ESAS) fatigue	severity	FACT-F and Quality of Life Questionnaire (QLQ)-C30 and	ESAS		ESAS fatigue severity		Fatigue Questionnaire (FQ - Norwegian)
	Intervention	Radiation Therapy (RT)		Adjuvant RT (27–120	days post surgery)			Chemo and/or RT (for	rectal and SCLC patients)	Chemo or supportive	care									None
	Control	None mentioned		None				None		None										None fatigue d, n =120,
	Exclusion	Dementia, history of psychiatric disorder		Metastatic disease,	chemo with K1, 2nd cancer,	inflammatory disease, thyroid disease, history of depression, use of tranquilizers,	streroids, non- steroidal	None mentioned		Other cancer,	nematopoleuc stem-cell	transplantation, taking growth	factors							Breast cancer recurrence, other cancer (except melanoma, ovarian
	Sample	<i>n</i> = 15		<i>n</i> = 41				n = 18		Timepoint	(1)1=34 (23 Male,	II Female),	T2=28							Timepoint 1 (T1)=302, Timepoint 2
	Cancer	Uterine, 86% stage I, post hysterectomy		l	sресіпеd			l .	cell lung cancer (SCLC), NSCLC (stages IIIb–IV)		year of diagnosis									Breast, stages II and III
	Authors	Ahlberg et al., 2004	rain .	B ≱ a	ot al., v 5001 w	<i>mun</i> . Autho	r man	Olsoget	 t; avai ; ie	Panjuget	on P	MC	2013	Aug	ust 01.					Reinerts en et al., 2011

						Longitudinal	inal Studies				
Authors	Cancer	Sample	Exclusion	Control	Intervention	Measure	Data Collection	Marker	p value	Association	Covariates
		(T2)=236 (175 eliminated for	cancer in situ), depression in T1					IL1β mRNA	0.59–0.92, subset = 0.57– 0.8		
		reporting						IL1Brs1694 4 (A/G)	Not given		
		only in 1						IL6Rrs4129 267 (C/T)			
Б		point).						IL6Rrs4845 617 (A/G)			
Brain		Fatigued, n=55						IL6Rrs2228 145 (A/C)			
Beha								IL6Rrs1800 795 (G/C)			
av Im								CRPrs2794 521 (C/T)			
mur								CRPrs3091 244 (A/G/T)			
Vallarce	<u> </u>	n = 72	Do not meet COG/	None	Chemo + dexamethaso	Fatigue Symptom	DNA from blood pre	AHSG	None given	Expression not associated with	Ethnicity
et al.,tr 2010			St Jude low and standard risk		ne	Inventory (FSI), pediatric and parent	dexamethasone, 1, 2, 4, 8 hours post oral	IL6 G17AC		fatigue	
manı	using St Jude & Children's Oncology		criteria			versions	dexamethasone dose	IL6 C634G			
ıscrij								POLDIP3			
Wratter et al. 2004 2004 ui əldəli isve MA ni əldəli isve 1004 i	Breast, early stage	<i>n</i> = 52	Severe current illness, history of major breast surgery, history of breast irradiation treatment, no metastatic disease	None mentioned	Adjuvant RT post surgery	FACT-F	Blood before RT, weekly during RT, then 2 & 6 weeks post RT	Lymphocyte	Baseline and week 5 (p<0.01)	Decreased in both fatigued and nonfatigued groups from baseline to week 5	Body mass index (BMI)
201						Cross-secti	Cross-sectional Studies				
Authors	Cancer	Sample	Exclusion	Control	Intervention	Measure	Data Collection	Marker	p value	Association	Covariates
Alexand	Breast, stages 1–IIb	n = 60	Pregnancy, other	n = 104 (non-	Survivors, 3 months - 2	Brief Fatigue Scale	Peripheral Blood	Neutrophils	0.51	No association with fatigue	None mentioned
.1. 2006 2006		(rangne)	disease, confusion,	iaugue u)	year post printary therapy	(Brs)		Lymphocyte	0.25		
			demenua					Monocyte	0.052	Trended towards significant association with fatigue	
								Eosinophil	0.051		
Bower et	Breast	n = 19	Recurrence of	n = 18 (non-	Survivors recruited	RAND SF36 fatigue	Peripheral blood, fasting	White blood cells (WBC)	Not given	No association with fatigue	Age, income,
al., 2003		(rangned)	oreast cancer,	laugueu)	1994–1997	scare grouped subjects to fatigued		Granulocyte			eunicity, bivit,
								Monocyte			
							Lymphocytes extracted	CD8+T lymphocyte	0.124	31% more in fatigued group	
							Irom tasung biood	CD38 T lymphocyte	Not given	No association with fatigue	

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						Longitudi	Longitudinal Studies				
Authors	Cancer	Samule	Exclusion	Control	Intervention	Measure	Data Collection	Marker	n value	Association	Covariates
		audiuma	history of immuno			ond non fotiminad	1		A .mm.	TO THE COURT	depressed mood
			disease			during 2 assessments		HLA-DR T lymphocyte			cancer treatment
							Serum from fasting blood	L-1ra	0.253		
Bower et	Breast, stages 0–IIIA	n = 103	Neurologic or	None mentioned	Completed primary	FSI	Plasma	IL1ra	>0.70	No association with fatigue	Exposure to RT, age,
al., 2011b			mmune disease, smokers		months and not started			C-reactive Protein (CRP)			ume since reaument completion, BMI
Brain E					on endocrine treatment			Soluble TNF-RII		Especially with patients receiving therapy other than chemo	
eta Collado	Breast (stages 0,I, II)	n = 32	Cancer recurrence,	n = 18 (non-	Survivors, 1–5 years	SF36 - vitality scale	Lymphocytes taken in	CD8+T lymphocyte	>0.10	No difference between fatigued and	Age, BMI, time since
Hidalgo et al.mgo		(fatigued), < 50 score	immune disease, immunosuppr	tatigue d), score >70 in SF-36	post diagnosis	grouped subjects to fatigued and non-	the morning	CD18 B cells		non-fatigued groups	treatment, treatment mode, depressive
<i>mun</i> . Aı 900 00		with SF-36 fatigue	essant use, psychiatric disease, smokers, alcohol	fatigue scale		fatigued during 2–3 assessments		CD3-/CD16+/C D56+ natural killer (NK) cells			symptom scores
uthor		scale	use, SF-36 fatigue score 50–70,				Plasma taken in the	IL-6	Not given	Plasma levels did not reach	
manusc			advanced cancer stage, >5 years post treatment				morning	TNF-rII		significance	
Collagio Hidako et al.av	Breast (stage 0,I, II)	n = 33 (fatigued), SF36	Cancer recurrence, immune disease, on immunosuppressant	n = 14 (non- fatigue d), SF36 fatigue score	Survivors, 1–5 years post diagnosis	Multidimensional FSI	DNA from leukocytes extracted from peripheral blood	IL/1B511cytosine	0.052	Not significantly associated with fatigue after controlling for depression	Age, ethnicity, menopausal status, BMI, depressive
able i		score <55		0/<			Plasma	IL1ra	0.074	Marginally higher in fatigued group	symptoms, cancer treatment
MC 2013 Au For all Mg u MC 2013 Au Minter and a minter a minter and a	Various, incurable metastatic or locally advanced cancer (multicenter, international study, 16 centers)	n = 741 (324 with severe fatigue)	Physical/cognitive impairment, language problems	n = 417 with no severe fatigue	May be undergoing palliative anticancer treatment	QLQ-C30 (multi- language version): 3- item fatigue subscale	Whole blood drawn within 72 hours after obtaining questionnaire responses.	CRP	Not given	No association with fatigue	Age, BMI, disease stage
Orre	Testicular	n = 92	Mental retardation,	n = 191	Survivors, recruited	FQ (Norwegian)	Plasma drawn from	IL-6	0.835	No association with fatigue	Age
al., 2009		(chronic fatigue)	extragonadal germ cell malignancy	(without chronic fatigue)	from 1980–1998 database, 18–75 years		0800-12 noon; subjects allowed light breakfast	sTNF-RI	0.321		
			except skin), removal of non- affected testicle related to benign condition, concurrent infection or inflammation		pio			Neopterin	0.390		
Orre et	Breast (stage II, III,	n = 299	Cancer recurrence,	None	Survivors, treated from	FQ (Norwegian)	Blood drawn from	IL-6	0.76	Beta = -0.015	Age, educational level
al., 2011	unclassinable).		>/2 years old, other cancer except		adjuvant RT;		oyoo-12 noon; subjects allowed light breakfast	sTNF-RI	0.713,	Beta = 0.019	
			basal cell cancer or cancer in situ, prior		subsample of Reinersten et al., 2011			Neopterin	0.85	Beta = 0.009	

_											
	Covariates		Age, gender, time since treatment completion, hemoglobin	Sex, age, diagnosis,	Charlson comorbidity	index, and concurrent pharmcological	treatment	Type of adjuvant therapy, mood, network support, satisfaction, cortisol, perceived stress, optimism	Age, gender, BMI,	cancer site, previous cancer treatment,	comorbid disease, pain score, sleep disorder, dyspnea
	Association	Beta = 0.71	$R^2 = 0.11, \beta = 0.04$	$\beta = 8.8$	$\beta=10.78$	$\beta = 5.37$	$\beta = 4.6$	r = 0.07 - 0.14	I = 0.12	I = 0.05	Not given
	p value	0.183	0.78	>0.05				Not significant	0.29	0.67	0.75
	Marker	L-1ra	Lymphocytes	IL1β	IL-6	8-TI	TNF-α	TNF-a	IL-6	TNF-α	WBC
Longitudinal Studies	Data Collection		Retrospective extraction of clinical results (peripheral blood)		within 4 hours of collection			Blood collected 9–33 days post surgery (lumpectomy or mastectomy) before adjuvant therapy		tent	between 9–11 am
Longitudi	Measure		Functional Assessment of Chronic Illness Therapy-Fatigue (FACIT-F) – 2 items, I item from Hamilton Depression scale	Brief Fatigue	inventory (BF1)			Revised Piper Fatigue Scale (rPFS)	BFI – Korean version		
	Intervention	& Landmark-Høyvik H et al., 2009	None	None				Adjuvant chemo + RT	None		
	Control		None	None mentioned				None mentioned	Mild versus	moderate versus severe	categones of fatigue
	Exclusion	surgery for contralateral breast cancer stage I with no adjuvant therapy	None mentioned	Less than 18 years	or age			Psychiatric disorder, dementia, history or current substance abuse, thyroid issues, immune disorders, immunosuppr ession, other cancer except noninvasive type	Cognitive	impairment, chemo/RT to treat	active cancer, hematologic malignancies, use of psychostimula nt, h/o psychiatric disease, fever, use of antibiotics and antiepileptics, high TSH
	Sample		n = 44	n = 83				n = 44	06 = u	(mild=23, moderate	= 30, severe =37)
	Cancer		NSCLC, Stages IIIb and IV		3–11/2007			Breast, stage 0–IIIa			survival time is less than 6 months
	Authors		Paddison et al., 2009	Scheege	et al.mu	un. A	vutho	r kn anuscript; available in F 008 0 c t o N	Kwalk	- 2013 -;;	3 August 01.