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Blood donation from healthy donors

CD4+CD25- cell FACS sorting

Cell culture

Human CD4 effector T cell

**Highlights**

- VIMP is temporally upregulated after TCR stimulation in human CD4 effector T cells
- VIMP inhibits cytokine expression in human CD4 effector T cells
- VIMP inhibits cytokine expression via the NFATC2/Ca^{2+} signaling pathway
- VIMP inhibits cytokine expression by controlling E2F5 expression

Christophe M. Capelle, Ni Zeng, Egle Danileviciute, Sabrina Freitas Rodrigues, Markus Ollert, Rudi Balling, Feng Q. He

feng.he@lih.lu

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Identification of VIMP as a gene inhibiting cytokine production in human CD4+ effector T cells

Christophe M. Capelle,1,2 Ni Zeng,1 Egle Danileviciute,1,3 Sabrina Freitas Rodrigues,3 Markus Ollert,1,4 Rudi Balling,3 and Feng Q. He1,3,5,6,*

SUMMARY
Many players regulating the CD4+ T cell-mediated inflammatory response have already been identified. However, the critical nodes that constitute the regulatory and signaling networks underlying CD4 T cell responses are still missing. Using a correlation-network-guided approach, here we identified VIMP (VCP-interacting membrane protein), one of the 25 genes encoding selenoproteins in humans, as a gene regulating the effector functions of human CD4 T cells, especially production of several cytokines including IL2 and CSF2. We identified VIMP as an endogenous inhibitor of cytokine production in CD4 effector T cells via both the E2F5 transcription regulatory pathway and the Ca2+/NFATC2 signaling pathway. Our work not only indicates that VIMP might be a promising therapeutic target for various inflammation-associated diseases but also shows that our network-guided approach can significantly aid in predicting new functions of the genes of interest.

INTRODUCTION
CD4+ T cells represent a major subset of immune cells that are crucial for mounting and regulating an adequate immune response. However, during many infectious and complex chronic diseases, those T cells are dysregulated, either having an impaired responsive capacity or causing adverse effects through self-recognition and/or overactivation. Therefore, rebalancing the CD4+ T cell-mediated inflammatory response has been essential for the design of therapeutic options for those diseases (Zhu and Paul, 2010a). Although many players regulating the inflammatory response, cytokine production, and differentiation of CD4+ T cells have already been identified in the past (Brownlie and Zamoyska, 2013; Rodriguez-Jorge et al., 2019; Saez-Rodriguez et al., 2007; Zhu and Paul, 2010b), a thorough understanding of the regulatory and signaling networks governing inflammatory cytokine production in T cells is still missing. The gap is not only attributable to the long-standing nature of traditional trial-and-error experimental procedures but also to the lack of reliable high-throughput computational prediction.

VIMP, also known as NCBI: Selenoprotein S (SELS), SELENOS, TANIS, or SEPS1, is one of the only 25 genes encoding the 21st amino acid selenocysteine in humans (Schomburg, 2011). Located in the endoplasmic reticulum (ER) membrane, VIMP is mainly known as an important component of the ER-associated degradation (ERAD) complex (Kim et al., 2007; Qin et al., 2016) and physically binds to several ER membrane proteins (Lee et al., 2015; Ye et al., 2005). VIMP plays a role in mediating retro-translocation of misfolded proteins from the ER lumen to the cytosol, where the ubiquitin-dependent proteasomal degradation takes place (Ye et al., 2004). Genome-wide association studies have shown that polymorphisms in the promoter region of VIMP are linked to a wide spectrum of common complex diseases, including cardiovascular disease (Alanne et al., 2007), diabetes (Karlsson et al., 2004; Olsson et al., 2011), cancer (Meplan et al., 2010; Shibata et al., 2009; Sutherland et al., 2010), sepsis (He et al., 2014), and autoimmune diseases (Santos et al., 2014; Seiderer et al., 2007), in which activation of the immune system is believed to be dysregulated (Kuchroo et al., 2012).

Meanwhile, dysfunction of the ER and the unfolded protein response causes intestinal inflammatory diseases in several murine models (McGuckin et al., 2010). Additionally, a reduced expression of VIMP causes an increased expression of inflammatory cytokines, such as NCBI: IL6, IL1β, and TNFα in macrophages.
(Curran et al., 2005), as well as IL1β and IL6 expression in astrocytes (Fradejas et al., 2011). However, other studies did not show significant association between VIMP and the examined human inflammatory diseases (Martinez et al., 2008). This controversy underlines the necessity for a better understanding of how VIMP contributes to the pathogenesis of some inflammatory diseases, i.e., through which cell types and which molecular pathways VIMP contributes to the observed dysregulated inflammatory responses. Therefore, we sought to investigate whether and how VIMP plays a role in relevant specific immune cells, e.g., CD4+ T cells, a key subset of immune cells orchestrating different types of immune responses and being heavily involved in different complex diseases, as well as infectious diseases, such as COVID-19 (Braun et al., 2020; Mathew et al., 2020).

We have previously developed a correlation-network-guided approach, based on the guilt-by-association theory (Beyer et al., 2007; Gillis and Pavlidis, 2011; Oliver, 2000), to predict novel key genes of a given biological process or function and have successfully applied it to human CD4+CD25highCD127low regulatory T cells (Tregs) (Danileviciute et al., 2019; He et al., 2012). Here, we extended the strategy to human CD4+ effector T cells (Teffs) that were derived and expanded from sorted CD4+CD25 T cells by coculturing with EBV-transformed B cells and were able to predict that VIMP might play an important role in regulating the effector responses of Teffs. Combining both the network analysis and experimental verification, we identify VIMP as a previously unreported vital endogenous inhibitor of cytokine production in human CD4+ Teffs and reveal the molecular mechanisms through which VIMP regulates CD4+ Teff responses.

RESULTS

VIMP is temporally upregulated following TCR stimulation in Teffs

Using our previously reported high-time-resolution (HTR) time-series transcriptome data of Tregs and Teffs following TCR (T cell receptor) stimulation in the first 6 h (He et al., 2012), we observed that the transcript level of VIMP in Teffs temporally peaked within 2–3 h following stimulation, which was followed by a gradual decrease (Figure 1A). In contrast, the VIMP mRNA level was kept almost constant in Tregs during the first 6 h.
**Selected enrichment list:**
ER components: 1.7E-7
Known VIMP binding proteins: 2.0E-4
TCR signaling pathway: 1.2E-3

**ER stress genes**

**Cytokine genes**

**TCR genes**

**Cytokine mRNA expression**

**IL-2 secretion**

**Cytokine secretion**

**VIMP protein quantification**
following TCR stimulation (Figure 1A), indicating a possible specific role for VIMP in Teffs. Our quantitative real-time PCR (qPCR) results validated the transitionally elevated expression of the VIMP transcript in Teffs isolated from different healthy donors (Figure 1B). We also observed a correlation over time between the transcription levels of NCBI: VIMP, IL2, IL13, and CSF2 (GM-CSF) following TCR stimulation, indicating a potential regulatory relationship between VIMP and some of the cytokines in Teffs (Figure 1B). By flow cytometry (Figure 1C), we confirmed the gradual upregulation of VIMP protein expression in the first 5 h following TCR stimulation. In summary, both mRNA and protein expression of VIMP were upregulated following TCR stimulation, which was correlated with the expression of several examined cytokines, indicating a potential role of VIMP in regulating Teff responses.

**VIMP inhibition upregulates cytokine expression in Teffs**

The upregulation of VIMP and its correlation to cytokine expression encouraged us to further investigate VIMP’s potential role in CD4 T cell responses. We and others have previously shown that the enriched pathways, processes or functions among the genes surrounding a given hub gene in the correlation network might give valuable indications on potential new functions of the given hub gene (Danileviciute et al., 2019; He et al., 2012). Therefore, we used our correlation network-guided approach to predict the potential functions of VIMP by identifying the enriched pathways among the genes that are linked to VIMP within the subnetwork of the Teff correlation network, which was extracted from our published HTR datasets and networks (He et al., 2012) (Figure 2A).

Consistent with its known function and its localization in the ER membrane, the genes surrounding VIMP in the correlation network were significantly enriched for ER components (p value = 1.7 x 10^-7, cumulative binomial distribution) (Figure 2A). Furthermore, 3 of the 10 experimentally validated VIMP-binding partners found in the literature in other cellular types are directly linked to VIMP in the Teff correlation network (p value = 2.0 x 10^-4, http://string-db.org [Szklarczyk et al., 2019]), indicating the reliability of our method. Surprisingly, the pathway enrichment analysis shows that the genes linked to VIMP are significantly enriched for components involved in the TCR signaling pathway (p value = 1.2 x 10^-3, cumulative binomial distribution) (Figure 2A), suggesting a potential role of VIMP in the Teff response according to our network-based analysis strategy (Danileviciute et al., 2019; He et al., 2012). However, the genes linked to the hub gene of interests in the correlation network could follow at least two scenarios (He and Ollert, 2016; Langfelder and Horvath, 2008; van Dam et al., 2018). First, those genes could be co-regulated by chance with the hub gene and perform independent functions. Second, those genes could be co-expressed with the hub gene and play related roles in the same pathways to coordinate cellular resources for a particular function or purpose under certain conditions. We will test these possibilities in this work.

To systematically assess whether and how VIMP controls the inflammatory response of Teffs, we performed a transcriptome analysis of CD4 Teffs isolated from peripheral blood mononuclear cells (PBMCs) of three healthy donors that were subjected to a specific-small interfering RNA (siRNA) knockdown of VIMP
Considering that VIMP regulates the expression of cytokines, e.g., IL2 in CD4 Teffs via VIMP, which is in line with our knockdown results. As VIMP has reported functions in ER stress, we first checked the ER-stress responsive genes in the transcriptomic data of the Teffs transfected with si_VIMP versus that treated with control siRNA (si_NS). By perturbing the expression of VIMP, we expected a change in the expression of some ER-stress responsive genes. Nonetheless, our transcriptome data of Teffs with VIMP partial knockdown did not show any significant change in mRNA expression of those genes (e.g., NCBI: CHOP [DDIT3], GRP78 [HSPA5], EDEM1, DNAJC3 [PSIPK], and DNAJB9 [ERdj4] [Lee et al., 2003]) (Figure 2D). Only the expression of the ER-stress regulator XBP1 (Yoshida et al., 2001) was significantly but modestly decreased. Indeed, data from intestinal epithelial cells show that VIMP is only a marker, but not a regulator, of ER stress (Speckmann et al., 2014). That shows that the direct involvement of VIMP in ER stress might not be ubiquitous to all cell types. Therefore, we ruled out the possibility that VIMP directly regulates the expression of the ER-stress responsive genes, indicating other roles of VIMP in modulating the Teff responses.

Considering that the TCR signaling pathways were significantly enriched in the VIMP correlation network, we further analyzed the genes related to the TCR signaling pathway in Teffs after VIMP knockdown. Notably, we found 13 significantly upregulated genes involved in the TCR signaling, including several cytokines, namely, NCBI: IL2, IL4, CSF2, and IFNG (refer to https://www.genome.jp/kegg-bin/show_pathway?hsa04660) in the microarray datasets of the Teffs, although only subjected to a partial knockdown of VIMP (Figure 2E). Moreover, transcripts of the key TCR-related signaling molecules, such as NCBI: GRAP2, ZAP70, RASGRP1, and RAF1, were significantly affected (Figure 2E). With the observation in mind that VIMP and the TCR signaling-related genes were directly linked in our HTR correlation network (Figure 2A), this effect of the siRNA perturbation was not unexpected. Our results suggest that VIMP negatively regulates the expression of specific cytokines and influences the expression of important components of the TCR signaling pathway.

To further confirm whether VIMP regulates cytokine expression in Teffs, using PBMC of independent donors we measured the cytokine mRNA expression by qPCR and the secreted cytokines of Teffs that were exposed to a VIMP knockdown. Indeed, NCBI: IL2, IL21, and CSF2 mRNA were significantly upregulated in stimulated Teffs transfected with si_VIMP, compared with control Teffs (with si_NS) (Figure 2F). This observation was further consolidated by increased IL2, IL21, and GM-CSF protein levels in the culture media of stimulated Teffs transfected with si_VIMP, compared with that treated with control scrambled siRNA (Figure 2G). Furthermore, the VIMP knockdown also significantly promoted T cell proliferation as indicated by both carboxyfluorescein succinimidyl ester (CFSE) peak shifting and Teff cell number counting experiments (Figure 2H and 2I). As IL2 concentration was already significantly higher at 3 h following stimulation (Figure 2G) and no cell division was expected, the enhanced IL2 secretion following VIMP knockdown was not simply caused by more Teffs. All the analyses were done under the precondition that VIMP protein was successfully silenced (Figure 2J). In short, VIMP negatively regulates the expression of several cytokines in Teffs following stimulation.

Considering that VIMP encodes selenocysteine, thus requiring selenium (Se) for its protein synthesis, and the fact that a relatively low concentration of Se was used in our T cell media (IMDM, ~0.066 μM), we next supplemented sodium selenite to the T-cell culture media to the range of physiological concentrations (~1 μM) (Rauhamaa et al., 2008; Safaralizadeh et al., 2005; Stranges et al., 2011). In line with the reported inverse association between Se status and inflammatory bowel diseases (Kudva et al., 2015), increasing the concentration of Se in the media generated a dose-dependent suppressive effect on IL2 production of sorted CD4 Teffs following TCR stimulation (Figures S1A and S1B). Meanwhile, increasing Se concentration to a physiological concentration upregulated the VIMP expression of stimulated CD4 Teffs in three of five tested donors (Figures S1C and S1D). These results again indicate that Se, at least partially, negatively regulates the expression of cytokines, e.g., IL2 in CD4 Teffs via VIMP, which is in line with our knockdown results.

**VIMP controls cytokine expression via the transcription factor E2F5**

Next, we aimed to identify any (co-)transcription factors (TFs), whose expression were significantly affected after silencing VIMP, as they often serve as the key components orchestrating the activity of the relevant pathways. Through a systematic analysis of all the known mammalian TFs or co-factors (Zhang et al., 2012) in our microarray
datasets, NCBI: E2F5 was found to be the most significantly upregulated TF, following a partial VIMP knockdown (Figure 3A). Conversely, NCBI: RNF14 (ring finger protein 14) was the most downregulated cofactor together with the downregulated TFs NCBI: CEBPG (CCAAT enhancer binding protein gamma), NCBI: ZBTB20 (zinc finger and BTB domain containing 20), and NCBI: IRX3 (Iroquois homeobox 3) (Figure 3A). We further confirmed the expression change of these (co-)TFs by qPCR in independent healthy donors (Figure 3B).

E2F5 has been reported to be a downstream target of IL-2 in an immortalized human T cell line (Brennan et al., 1997). But to our knowledge, there are no reports yet of E2F5 sitting at the upstream pathways regulating inflammatory responses, especially cytokine production. Nevertheless, being the most significantly upregulated TF following a partial knockdown of VIMP, we assumed that E2F5 might be an important component in the regulatory pathway through which VIMP regulates the Teff inflammatory response.

Therefore, we decided to investigate whether VIMP controls the cytokine expression by negatively regulating E2F5 expression in stimulated Teffs. To examine this hypothesis, we silenced VIMP alone or in combination with E2F5 and measured the expression of selected cytokines by qPCR. In addition to the reduced expression of VIMP, the upregulation of E2F5 expression that was driven by VIMP knockdown was abolished in the VIMP and E2F5 double knockdown Teffs (Figure 3C). Silencing VIMP alone upregulated IL2 expression in stimulated Teffs, whereas a dual knockdown of VIMP and E2F5 suppressed the surge of IL2 caused by VIMP knockdown alone (Figure 3C). Even though E2F5 is a general regulator of transcription, we did not observe any effect of E2F5 knockdown on genes that are not directly involved in Teff inflammatory response, such as NCBI: CTLA4 (Figure 3C). This excluded a generalized effect of E2F5 on the transcriptional regulation in Teffs. In brief, our data support the fact that VIMP regulates the expression of inflammatory cytokines, i.e., IL2, by restraining the expression of the TF E2F5 in Teffs.

**VIMP controls cytokine expression via the Ca²⁺/NFATC2 signaling pathway**

To further delineate VIMP’s regulatory pathways beyond the altered expression of individual TFs determined by the differential expression analysis of our microarray datasets, we applied the Ingenuity Pathway Analysis (IPA) to
map the up- or downregulated cytokine and TCR related genes into the known regulatory network structures. We found that many of those differentially expressed genes are controlled by the expression change of the so-called hub genes NCBI: IL2, RAF1, IL21, and TNFSF11, as well as nuclear factor of activated T cells (NFAT) activity (Figure 4A). Although NFAT transcript expression was not significantly affected (Figure 4A), its activity was predicted to be increased by the IPA computational analysis. Meanwhile, we investigated the VIMP subnetwork in the Teff correlation network in more depth (Figure 2A). We found that genes for several components of NF-kB, NFAT, and MAPK signaling pathways were also directly linked to VIMP, indicating that those pathways might be involved in the regulation of the inflammatory response of Teffs by VIMP. To determine whether any of the relevant signaling pathways downstream of the TCR pathway that were suggested by the computational analysis are affected by VIMP expression, we quantitatively assessed the phosphorylation levels of up to 10 various signaling proteins by flow cytometry (Figure 4B). Canonical (NFKB1, p105, and p65) and non-canonical (NFKB2, p100, and RELB) NF-kB signaling pathways, as well as several MAP kinase sub-pathways (ERK1/2, p38, JNK1/2, and c-Jun) were not significantly affected in their phosphorylation levels (Figures S2A–S2E and S3). The phosphorylation level in one of the NFAT family members, NFATC1, was also not significantly affected by VIMP knockdown in stimulated Teffs (Figures S2F and S3). However, the phosphorylation level at the specific site Ser326 of another NFAT family member, NCBI: NFATC2 (also known as NFAT1), was significantly reduced even following a partial VIMP knockdown, as quantified by both flow cytometry and western blotting in Teffs isolated from different donors (Figures 4C–4E and S2G). Total NFAT1 protein remained unaffected by the partial VIMP knockdown (Figure S2H). In resting T cells, NFAT proteins are phosphorylated and reside in the cytoplasm (Okamura et al., 2000; Sharma et al., 2011). To be able to translocate to the nucleus and induce gene expression, NFAT is de-phosphorylated following the TCR signaling. As the NFAT activity is known to regulate IL2 expression in T cells (Chow et al., 1999), the observed downregulation of NFATC2 phosphorylation, following VIMP knockdown, demonstrated that the upregulation of IL2 expression was, at least in part, due to an increase in NFAT activity.

The distinguishable feature of NFAT is that it relies on Ca^{2+} influx and subsequent Ca^{2+}/calmodulin-dependent phosphatase calcineurin to become dephosphorylated and being able to translocate to the nucleus to induce gene expression (Hogan et al., 2003). Although VIMP has not yet been linked to the calcium signaling, other selenoproteins have been described to regulate the calcium signaling and homeostasis (Pitts and Hoffmann, 2018). We therefore further asked whether VIMP knockdown affects the calcium flux in Teffs and measured it by flow cytometry using the calcium indicator Indo-1. Indeed, the Teffs in which VIMP was silenced versus the control Teffs showed a significantly higher flux of Ca^{2+} ions (Figure 4F), further illustrating the increased NFATC2 activity and IL2 expression.

In summary, our data strongly suggest that VIMP inhibition upregulates the expression of cytokines, such as IL2, by two mechanisms at different levels (Figure 4G). On the transcription regulatory level, VIMP controls the expression of TF E2FS and multiple genes involved in the TCR signaling and the inflammatory response. On the signaling transduction level, VIMP knockdown modulates Teff responses by controlling the Ca^{2+} flux and the downstream NFATC2 de-phosphorylation.

**DISCUSSION**

So far many important components in the regulatory or signaling networks modulating the inflammatory responses of Teffs still remain elusive. With the development of systems medicine, researchers have
greater opportunities to use top-down approaches to objectively infer and identify novel key genes or proteins in the process of interest.

In this work, we have applied our previously published correlation network-guided strategy to predict new genes regulating the effector functions of CD4+CD25- Teff cells, i.e., cytokine production. We identified VIMP, encoding an ER membrane-associated selenoprotein, as a previously unrecognized negative regulatory gene of the Teff response. VIMP is best known for its critical functions in ER stress, which was demonstrated in some tested cell types. Our transcriptomic correlation network in Teffs also indicates that VIMP might be involved in ER-stress-related functions. However, as shown here, inhibiting the VIMP expression in primary human Teffs did not support the fact that VIMP is critical for the transcriptional regulation of ER-stress responsive genes in Teffs. Next, the correlation network navigated us to check the TCR signaling pathways. As demonstrated in different layers, VIMP indeed substantially regulated the expression of several inflammatory cytokines, especially IL2, in Teffs. We next investigated the VIMP regulatory mechanisms using primary human Teffs isolated from different healthy donors, the most clinically relevant available materials. Combining the analysis of time-series correlation network with knockdown-based regulatory networks, we further predicted that VIMP might go through the NFAT signaling pathway, or MAP kinase or NFKB signaling pathways to mediate the effector functions of Teffs. After testing those signaling pathways one by one, we finally pinpointed that VIMP inhibition enhances cytokine production of Teffs via the NFATC2 signaling pathway. The involvement of the NFAT signaling pathway was further backed by the influence of VIMP inhibition on calcium (Ca²⁺ influx, which is vital to the activation of the NFAT signaling pathway). Coincidently, Joost and colleagues have recently reported the co-expression sequencing analysis, indicating from another angle that our conclusion might hold true (Joost et al., 2016). We have also shown that E2F5 plays a significant role in the VIMP-mediated regulation of the Teff IL2 expression. However, whether the E2F5 pathway and the Ca²⁺/NFATC2 signaling controls VIMP-mediated IL2 expression in a sequential manner or in parallel requires further investigation. Although the published association studies have already shown that the VIMP expression levels and/or SNPs are correlated with the risk of several types of diseases, it remains unsolved whether VIMP deficiency can regulate the effector functions of Teffs in vivo.

In our TF-focused analysis, we identified not only E2F5 as the most upregulated TF but also several downregulated TF genes, following VIMP knockdown. Among those downregulated ones, RNF14 (ring finger protein 14), a less characterized gene, represented the most significantly downregulated co-factor, attributable to VIMP knockdown in Teffs. Although very limited, a published report shows that RNF14 modulates the expression of inflammatory and mitochondria-related genes in a murine myoblast cell line (Ingham et al., 2014). Another downregulated TF ZBTB20, originally studied in human dendritic cells (Zhang et al., 2001), and later in myeloid cells (Liu et al., 2013) and B cells (Zhu et al., 2018), has been shown to regulate their effector functions and differentiation. The Iroquois homeobox 3 (IRX3) has been recently linked to human CDB T cell survival and fate determination in vitro (Persengiev, 2017). Although there is no direct evidence of CEBPG being involved in the regulation of cytokine expression in CD4 effector T cells, other C/EBP protein family members have been shown to act as negative regulators in the production of inflammatory cytokines (Berberich-Siebelt et al., 2000; Tanaka et al., 2014). Therefore, those TFs might deserve further investigation.

Selenoproteins fully rely on selenium for their biosynthesis and function. Dietary selenium supplementation in mice has been shown to increase the biosynthesis of several selenoproteins including SELS/VIMP (Stoedter et al., 2010; Tsuji et al., 2015) and to affect the expression of several inflammatory cytokines (Beck et al., 2001; Hao et al., 2016; Stoedter et al., 2010; Vunta et al., 2007; Zhu et al., 2017). Dietary selenium supplementation has further been linked to alleviate several complex and multifactorial diseases (Duffield-Lillico et al., 2003; Kim et al., 1999, 2000; Kudva et al., 2015). On the other hand, selenium deficiency might affect the synthesis of multiple selenoproteins in mice, resulting in an increased pathology from viral or bacterial infections (Beck et al., 2001; Gao et al., 2016). In our media (complete IMDM) for short-term T cell culture, the Se concentration (0.066 μM) was around 15 times lower than in human sera (~1 μM) (Rauhamaa et al., 2008; Safaralizadeh et al., 2005; Stranges et al., 2011). Although the Se concentration used in our media was low, our western blotting results (Figure 4C) have demonstrated that the Se concentration was not yet a limiting factor for VIMP protein synthesis during the tested period of up to 24 h following stimulation, as the protein expression of VIMP still increased following TCR stimulation. In the VIMP-knockdown T cells,
where the VIMP protein synthesis was further reduced, the low concentration of Se in the media was thus not a concern. Therefore, our conclusion derived from IMDM media with a VIMP-knockdown approach is reliable. Last but not least, increasing Se concentration showed a dose-dependent suppressive effect on IL2 production (Figures S1A and S1B). Following Se supplementation, the majority of the tested donors exhibited enhanced expression of VIMP in CD4 Teffs (Figure S1D). For the other donors, already having a high level of VIMP expression (Figure S1D), Se supplementation cannot further increase the expression of VIMP, but still inhibited cytokine production, possibly via enhancing the expression of the other selenoproteins as an alternative pathway. These observations indicate that at least for some patients with VIMP deficiency, Se supplementation would show beneficial values in suppressing pro-inflammatory cytokine responses of CD4 T cells.

Interestingly, the immune system presents a sexual dimorphism (Klein and Flanagan, 2016), where females appear to have a stronger humoral and cellular immune response in general, making them more resistant to infectious diseases (vom Steeg and Klein, 2016), nevertheless, more susceptible to autoimmune diseases (Angum et al., 2020; Jacobson et al., 1997). CD4 T cells, the focus of this study and the central orchestrators of immune responses, also show a differential sex-specific regulation (Afshan et al., 2012; Aldridge et al., 2018; Klein and Flanagan, 2016). Multiple factors on the genetic (Bianchi et al., 2012; Cacciari et al., 1981; Kocar et al., 2000), hormonal (Straub, 2007), or environmental level (Jensen et al., 2016; Kawai et al., 2010) have been shown to regulate sex-specific effects in immune responses. It is worth noting that selenium also displays intriguing sex-specific differences in regard to its metabolism (Seale et al., 2018), tissue distribution (Pitts et al., 2015), and effects in several physiological and pathological conditions, including immune-associated diseases (Li et al., 2020; Lu et al., 2019; Riese et al., 2006; Schomburg and Schweizer, 2009; Stoedter et al., 2010; Waters et al., 2004). Excitingly, the expression of VIMP increases following selenium supplementation in the liver of male mice, whereas in female mice VIMP expression only reaches a maximum after LPS challenge to induce an acute immune response (Stoedter et al., 2010). In regard to our data, this leads us to hypothesize that selenium supplementation and its potential sex-specific effects on VIMP expression might also result in a gender-biased effect on CD4 T cells.

Overall, using both hypothesis-free top-down computational analyses and bottom-up experimental methods, we have shown a regulatory role for the selenoprotein VIMP in controlling cytokine expression in CD4+CD25+ Teffs via several signaling pathways and transcriptional regulatory pathways. The same strategy should be generally extendable to other cell types in assisting the prediction and discovery of novel functions of any other genes of importance. In summary, our data identified an unrecognized critical regulatory role of the selenoprotein S (SELS/VIMP) in the inflammatory responses of human CD4+ Teffs. Our observation provides a viable insight into how dietary supplementation of selenium might mediate its effects on CD4+ Teffs and underscores the potential in therapeutically targeting VIMP in the treatment of various inflammatory and inflammation-related diseases.

Limitations of the study

Although we have successfully demonstrated an unrecognized role for VIMP in the regulation of CD4 T cytokine expression and the underlying mechanisms, our study still presents certain limitations. As aforementioned, selenium supplementation and immune cell responses display a sexual dimorphism. In this study, we are aware that the majority of healthy donors were male. However, due to ethic regulations, we could not specify the gender of each individual donor, making it impossible to determine a possible sex-specific effect of VIMP on the effector functions of CD4 T cells.

In addition, our data is based on primary human CD4 T cells expanded in vitro and do not take into account all the complex cellular regulatory mechanisms directly and indirectly acting on CD4 T cells present in vivo. Our work has shown an intrinsic role of VIMP on human CD4 effector T cells, but to better elucidate the importance of our findings in a disease context, Vimp-deficient mice could have been used, which, however, were not available in our laboratory. Even though the whole-body deficiency of some selenoproteins is embryonically lethal (Santesmasses et al., 2020), Vimp-deficient mice have been recently reported and mainly used to study the role of Vmp in muscle functions (Addinsall et al., 2018, 2020; Wright et al., 2017). Excitingly, in line with our notion in human CD4 T cells, the reduction of Vmp expression even in heterozygous mice has been shown to increase the expression of several inflammatory genes in fast-twitch skeletal muscles (Wright et al., 2017).
Resource availability

Lead contact
Further information and requests for different resources should be directed to and will be fulfilled by the lead contact, Feng He, Department of Infection and Immunity, Luxembourg Institute of Health, Esch-sur-Alzette, Luxembourg (feng.he@lih.lu).

Materials availability
The study did not generate any new unique materials.

Data and code availability
The whole-transcript microarray data have been deposited into Gene Expression Omnibus (GEO) repository with the access code https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE151266. Raw gel images from Figure 4C were deposited on Mendeley at https://doi.org/10.17632/6bd75yg6rp.1.

METHODS
All methods can be found in the accompanying Transparent methods supplemental file.

SUPPLEMENTAL INFORMATION
Supplemental information can be found online at https://doi.org/10.1016/j.isci.2021.102289.

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AUTHOR CONTRIBUTIONS
C.M.C., N.Z., E.D. designed and performed experiments. C.M.C. analyzed the data and wrote the manuscript. S.F.R. performed parts of the experiments. R.B., M.O., and F.Q.H. supervised the project. F.Q.H. designed the project, oversaw the whole project, and revised the manuscript. All the authors read and edited the manuscript.

DECLARATION OF INTERESTS
The authors declare that they have no conflict of interest.

INCLUSION AND DIVERSITY
We worked to ensure ethnic or other types of diversity in the recruitment of human subjects. The author list of this paper includes contributors from the location where the research was conducted who participated in the data collection, design, analysis, and/or interpretation of the work.

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Supplemental information

Identification of VIMP as a gene inhibiting cytokine production in human CD4+ effector T cells

Christophe M. Capelle, Ni Zeng, Egle Danileviciute, Sabrina Freitas Rodrigues, Markus Ollert, Rudi Balling, and Feng Q. He
Supplemental Figure S1. Selenium supplementation suppresses IL2 production in CD4 Teffs, Related to Figure 2. Human CD4+CD25- T cells sorted from healthy donors were unstimulated in normal IMDM complete media or stimulated for 24 hrs by soluble anti-CD3/CD28 antibodies either in IMDM complete media alone or supplemented with different concentration of sodium selenite (S5261, Sigma Aldrich). (A, C) The mRNA expression of IL2 (A) and VIMP (C) was quantified by qPCR and normalized to the housekeeping gene RPS9 and to that of the unstimulated samples of the given donor. (B, D) A “zooming-in” analysis of the two selected concentrations of Se for IL2 (B) and VIMP (D). Each dot represents one healthy donor. The donor ID was indicated for the concentration of interests. Data are mean± standard deviation (s.d.). The P-values are determined.
by a two-tailed paired Student’s t test. ns or unlabeled, non-significant; *P<=0.05, **P<=0.01 and ***P<=0.001.
Supplemental Figure S2. VIMP knockdown only affects the phosphorylation of NFATC2, not the other major signaling pathways downstream of the TCR, Related to Figure 4.

Phosphorylation of proteins involved in the major signaling pathways downstream of TCR signaling in Teffs, assessed by flow cytometry at different time points following PMA/ ionomycin stimulation. Before stimulation, the cells were transfected with specific siRNA against VIMP (si_VIMP) versus non-specific siRNA (si_NS) for 1 day. (G) Only pNFATC2 was significantly decreased by VIMP knockdown. The other measured targets remain no significant change (A-F, H). The fold change was calculated by normalizing the geometric mean (Geomean) of the fluorescence intensities of all the conditions to that of the unstimulated control knockdown condition. Data are mean± standard deviation (s.d.). The P-values are determined by a two-tailed paired Student’s t test over time including the data at different time points. ns or unlabeled, non-significant; *P<=0.05, **P<=0.01 and ***P<=0.001. All the graphs represent the pooled flow cytometry data for the fold change from 2-7 independent donors.
Supplemental Figure S3. VIMP knockdown does not affect other major signaling pathways downstream of the TCR, Related to Figure 4.

Representative histogram overlay for the phosphorylation of major signaling transduction proteins downstream of the TCR signaling in Teffs, assessed by flow cytometry at different time points following PMA/ionomycin stimulation. Before stimulation, the cells were first transfected with specific siRNA against VIMP (si_VIMP) versus non-specific siRNA (si_NS) for 1 day. No significant effects on the phosphorylation levels of MAPK (p38, ERK1/2, cJun, JNK1/2) pathways and canonical (p65, p105) or non-canonical (RELB, NFκB2) NFκB pathways during the first 120 min stimulation after siRNA knockdown in Teffs. The expression of total NFAT1 protein was also unaffected by VIMP knockdown. The numbers in x-axis indicate the geometric mean (Geomean) fluorescence intensity of the different proteins or phosphorylation sites. Data are mean± standard deviation (s.d.). The other measured targets remain no significant change (A-G). The P-values are determined by a two-tailed paired Student’s t test. ns or unlabeled, non-significant; *P<=0.05, **P<=0.01 and ***P<=0.001. All the graphs represent data from 2-7 independent donors.

Transparent Methods

Primary T cell isolation and culture

Buffy coats from healthy donors were provided by the Red Cross Luxembourg and the informed consent was obtained from each donor by the Red Cross Luxembourg. The T cell isolation and culture procedures have been described in our previous works (Danileviciute et al., 2019; He et al., 2012; Sawlekar et al., 2020). For the requirement of the STAR methods, we briefly described it here again. We added the RosetteSep™ Human CD4+ T cell Enrichment Cocktail (15062, Stemcell) to undiluted blood at a concentration of 50 µl/ml and incubated for 30 min at 4°C. The incubated samples were then diluted 2 times with FACS buffer (PBS + 2% FBS) and the CD4+ T cells were obtained following gradient centrifugation at 1200 g for 20 min, using Lympoprep (07801, StemCell) and SepMate™-50 tubes (85450, Stemcell). Before sorting, the CD4+ T cells were first stained with mouse monoclonal [RPA-T4] anti-human CD4 FITC (555346, BD Biosciences) (dilution 1:20), mouse monoclonal [M-A251] anti-human CD25 APC (555434, BD Biosciences) (dilution 1:20), and LIVE/DEAD® Fixable Near-IR Dead Cell Stain (L10119, Thermo Fisher Scientific) (dilution 1:500). Primary CD4 T cells (CD4+CD25) were then sorted on a BD FACSAria™ III cell sorter (BD Biosciences).

| Target     | Fluorochromes | Dilution | Company          | Clone  | Reference |
|------------|---------------|----------|------------------|--------|-----------|
| CD4        | FITC          | 1:20     | BD Biosciences   | RPA-T4 | 555346    |
| CD25       | APC           | 1:20     | BD Biosciences   | M-A251 | 555434    |
| Live/Dead  | Near Infra-Red| 1:500    | Thermo Fisher Scientific | N.A.   | L10119    |

Sorted human CD4+ T cells were cultured in IMDM (21980-032, Thermo Fisher Scientific) complete medium, supplemented with 10% heat-inactivated (56°C, 45 min) fetal bovine serum (FBS) (10500-064, Thermo Fisher Scientific), 1x Penicillin+Streptomycin (15070-063, Thermo Fisher Scientific), 1x MEM non-essential amino acids (M7145, Sigma-Aldrich) and 1x β-mercaptoethanol (21985-023, Thermo Fisher Scientific).
Thermo Fisher Scientific). Every seven days for a maximum of four weeks, Teffs were derived from isolated CD4+CD25- T cells by restimulating them with irradiated Epstein–Barr virus (EBV) transformed B-cells (EBV-B cells) (Probst-Kepper et al., 2009), at a 1:1 ratio to expand and maintain the culture. The EBV-B cells were irradiated in RS2000 X-Ray Biological Irradiator (Rad Source Technologies) for 30 min with a total of 90 Gy.

**Teff siRNA knockdown and stimulation**

Targeted gene’s expression was knocked-down in up to 5 x 10^6 cells using the P3 Primary Cell 4D-Nucleofector X Kit L (V4XP-3024, Lonza) with 90 µl P3 Primary cell solution and 100 pmol of corresponding si_RNA (resuspended in 10 ul RNase-free H2O): si_Non-Specific scrambled control siRNA (si_NS or si_CTRL) (SC-37007, Santa Cruz), si_VIMP/SELS (SI03053512, Qiagen), si_E2F5 (SI00030436, Qiagen). siRNA transfection by electroporation was performed in the Amaza 4D-Nucleofector™ X System (Lonza) according to the manufacturer’s recommended program for primary human T cells (with the program code EO-115). Following transfection, the Teffs were first transferred into a 12-well plate with pre-warmed complete IMDM medium and incubated at 37 °C for 24 hrs before being stimulated with 25 µl/ml of soluble antibodies (Immunocult™ Human CD3/CD28 T Cell Activator) (10971, StemCell), or 10 ng/ml PMA (Phorbol 12-myristate 13-acetate, P8139, Sigma-Aldrich) and 100 ng/ml Ionomycin (I0634, Sigma-Aldrich) or Dynabeads® Human T-Activator CD3/CD28 for T Cell Expansion and Activation (11131D, Thermo Fischer Scientific) (with 1:1 ratio between number of cells and beads) in a 24-well plate for different specified time periods.

**RNA extraction**

The RNeasy Mini Kit (74106, Qiagen) was employed for RNA extraction following the manufacturer’s instructions and including the digestion of genomic DNA with DNase I (79254, Qiagen). The cells were lysed in RLT buffer (79216, Qiagen), supplemented with 1% beta-Mercaptoethanol (63689, Sigma-Aldrich). NanoDrop 2000c Spectrophotometer (Thermo Fisher Scientific) was used to measure RNA concentration. For the microarray analysis, the quality of RNA was first checked by assessing the RNA integrity number (RIN) using the Agilent RNA 6000 Nano kit (5067-1511, Agilent) and the Agilent 2100 Bioanalyzer Automated Analysis System (Agilent), according to the manufacture’s protocol. Only the samples with RIN of 8.5 or higher were used in the further analysis.

**Microarray measurement and analysis**

The transcriptomic analysis of human effector T cells expanded from CD4+CD25- T cells isolated from the PBMCs of healthy donors were performed on the Affymetrix human gene 2.0 ST array at EMBL Genomics core facilities (Heidelberg, led by Dr. Benes Vladimir). The facility used 500 ng of total RNA in the protocol with the Ambion® WT Expression Kit (cat. 4411974) in order to obtain 10 µg of cRNA, which was then converted to ssDNA. 5.5 ug of ssDNA was labeled and fragmented using the WT Terminal Labeling, polyA and hyb Controls Kit (Affymetrix, cat. 901524). 3.75 ug of fragmented/labeled ssDNA (with hybridization controls) was hybridized to Affymetrix HuGene 2.0 Genechip at 45 °C for 16 hrs with rotation (60 rpm) and washed and stained on GeneChip Fluidics Stations 450 using GeneChip® Hybridization Wash and Stain Kit (Affymetrix, cat. 900720). Arrays were scanned using GeneChip Scanner 3000 7G with GeneChip Command Console software.
The expression signal at the exon level was summarized by the Affymetrix PLIER approach using the sketch approximation of quantile normalization with the option PM-GCBG (a GC content based background correction) using Affymetrix Expression Console v1.3.1.187. Before performing differential analysis, we first pre-processed the data with certain filtering steps. The filtering steps following the PLIER summary method included: 1) first removing any probeset whose cross-hyb type was not equal to 1; 2) removing any probeset corresponding to no identified gene or multiple genes according to the annotation (the file HuGene-2.0-st-v1.na33.2.hg19.transcript) and the library version r4 (May 23, 2012); 3), excluding the probesets with the average expression value in both groups (si_NS and si_VIMP) ≤ 2 times of the median value of the arrays (in our case, 2x the median was equal to the intensity value of 170); 4) if the mean intensity of the probesets in one group was higher, the number of absent calls among the three biological replicates should not be ≥ 1 in the group with higher mean intensity. To secure more robust analysis, we also analyzed the dataset using another model-based method (Berchtold et al., 2008; Weigand et al., 2012), i.e., RMA-sketch summary/normalization method (of note, the filtering steps mentioned above did not apply to the data resulted by the RMA-sketch summary method). We selected the probeset for further analysis only if the two-sided pair-wised T-test generated a P-value lower than 0.05 from the datasets summarized by both PLIER and RMA methods as demonstrated somewhere else (Weigand et al., 2012). To obtain a certain number of starting candidates, we lowered the threshold of the change fold to 1.2, which had to be recurrent in all the three donors, for our further analysis in consideration of both facts that VIMP is not a (co)transcription factor and the siRNA knockdown efficiency was not 100%. The database of mammalian transcription factors or cofactors, or chromatin remodeling factors was downloaded from the work of others (Zhang et al., 2012).

In this way, around 800 genes were significantly upregulated and around 550 genes were downregulated following VIMP knockdown, which were used for further analysis.

**Correlation network and IPA**

The Teff correlation network based on high-resolution time series datasets of Teffs was already calculated and constructed in our previous work (He et al., 2012) and we extracted the VIMP subnetwork for a deeper analysis in this work. Ingenuity Pathway Analysis (IPA) was used to reconstruct the regulatory network from the Ingenuity database following the instruction of provider (QIAGEN).

**cDNA synthesis**

The SuperScript™ IV First Strand Synthesis System (18091050, Thermo Fisher Scientific) was used for human cDNA synthesis using a maximum of 500 ng of RNA following the manufacturer's protocol. The master mix for the first step per sample including: 0.5 µl of 50 µM Oligo(dT)20 primers (18418020, Thermo Fisher Scientific), 0.5 µl of 0.09 U/µl Random Primers (48190011, Thermo Fisher Scientific), 1 µl of 10 mM dNTP mix (18427013, Thermo Fisher Scientific) and RNase free water for a final volume of 13 µl in 0.2 ml PCR Tube Strips (732-0098, Eppendorf). The C1000 Touch Thermal Cycler (Bio-Rad) or UNO96 HPL Thermal Cycler (VWR) were employed with the following program: 5 min at 65 °C, followed by 2 min at 4 °C. For the second reaction step, the reaction mix was accompanied with 40 U RNaseOUT™ Recombinant Ribonuclease Inhibitor (10777019, Thermo Fisher Scientific), 200 U SuperScript™ IV Reverse Transcriptase (18090050, Thermo Fisher Scientific), a final concentration of 5mM Dithiothreitol (DTT) (707265ML, Thermo Fisher Scientific)
and 1x SuperScript™ IV buffer to reach a final reaction volume of 20 µl. We used the following thermocycler program for the second step: 10 min at 50 °C, then 10 min at 80 °C and at 4 °C until further usage. The nuclease-free water was used to dilute the obtained cDNA 5 times with a final volume of 100 µl.

**Quantitative real-time PCR**

The quantitative real-time PCR (qPCR) reaction mix per sample enclosed: 5 µl of the LightCycler 480 SYBR Green I Master Mix (04707516001, Roche), 2.5 µl cDNA and 2.5 µl primers in a total reaction volume of 10 ul. The PCR reaction was performed in a LightCycler 480 (384) RT-PCR platform (Roche), using the LightCycler 480 Multiwell 384-well plates (04729749 001, Roche) sealed with the LC 480 Sealing Foil (04729757001, Roche). The program for qPCR used was as follows: 5 min at 95 °C; 45 cycles of (10 sec at 55 °C, 20 sec at 72 °C, 10 sec at 95 °C); melting curve (65-97 °C). The results were analyzed using the LightCycler 480 SW 1.5 software. Primers used for qPCR: RPS9 (QT0023989, Qiagen) as a reference gene, VIMP/SELS (QT00008169, Qiagen), IL2 (QT00015435, Qiagen), CSF2 (QT0000896, Qiagen), IL21 (QT00038612, Qiagen), CEBPG (QT00224357, Qiagen), E2F5 (QT00062965, Qiagen), IRX3 (QT00227934, Qiagen), RNF14 (QT00088291, Qiagen), ZBTB20 (QT00069776, Qiagen) and CTLA4 (QT01670550, Qiagen).

**Western blotting**

Novex™ WedgeWell 4-20% Tris-Glycine pre-casted gels (XP04202Box, Invitrogen) were used to run and separate proteins in the Novex™ Tris-Glycine SDS Running buffer (LC2675-4, Invitrogen). The proteins were then transferred (dry transfer) using an iBlot2™ Gel Transfer Device (IB21001, Invitrogen) and iBlot2™ PVDF stacks (IB24002, Invitrogen). Following the transfer, the membranes were blocked in 5% milk in PBS with 0.2% Tween20 (PBS-T) for 1 hr at room temperature with gentle shaking and incubated overnight at 4°C together with the primary antibodies, diluted in 5% BSA in PBS-T with 0.025% sodium azide. The next day, the membrane was washed three times (10 min each time) before and after incubation with secondary goat anti-rabbit HRP-coupled antibodies. The Amersham ECL Prime Western Blotting Detection Reagent (RN2232, GE Healthcare Life Sciences) was used to detect the proteins and the image of the membranes was visualized on the ECL Chemocam Imager (INTAS). If necessary, the contrast and brightness of the obtained whole gel pictures was adjusted using ImageJ. The signal intensity of the protein bands was quantified using ImageJ and normalized to that of the housekeeping gene GAPDH. For the quantification of phospho proteins, both the phospho and the pan protein were normalized to GAPDH, before normalizing the phospho protein to the total protein.

| Target          | Dilution | Company            | Clone   | Reference     |
|-----------------|----------|--------------------|---------|---------------|
| pNFATC2 (Ser326)| 1:100    | Sigma-Aldrich      | SAB4503945 |
| NFAT1           | 1:1000   | Cell Signaling     | D43B1   | 5861S         |
| VIMP            | 1:1000   | Sigma-Aldrich      | Polyclone | V6639       |
| GAPDH           | 1:200    | Santa Cruz Biotechnology | FL-335 | SC-25778     |

**Deposited gel data:**

[http://dx.doi.org/10.17632/6bd75yg6rp.1](http://dx.doi.org/10.17632/6bd75yg6rp.1)
Proliferation assay

The proliferation of the Teffs was assessed using the CellTrace™ CFSE cell proliferation kit (C34554, Invitrogen). The final concentration of 1 μM CFSE dye was used in our work. To label the cells, they were incubated for exactly 2 min and 45 sec at RT in the dark. To stop the reaction, 10 ml FBS was added and the cells were centrifuged at 200 g for 10 min. After washing the cells in IMDM medium, the cells were subjected to the siRNA knockdown and counted. 10⁵ Teff in a 96-well plate were used for each condition and stimulated for 2 days with a ratio of 1:1 of irradiated Epstein Barr Virus (EBV) B cells as previously described (He et al., 2012). After the stimulation, the cells were stained for living cells using LIVE/DEAD® Fixable Near-IR Dead Cell Stain (L10119, Thermo Fisher Scientific) (dilution 1:500) and acquired on a BD Fortessa™ analyzer. The data was analyzed in FlowJo 7.6.5.

Cytokine measurement by Mesoscale discovery (MSD) platform

The cell supernatant was collected after centrifugation of the cells (250 g, 10 min) and the selected list of secreted cytokines (CSF2, IL2, IL21) was measured in the undiluted cell culture medium using the MSD U-PLEX Human Biomarker group 1 kit (MSD, K15067L-1) and following the manufacturer’s instructions. MESA QuickPlex SQ 120 instrument was used to read the plate and the data was analyzed with the MSD Workbench software.

Cytokine measurement by Cytometric Bead Array (CBA)

The cell supernatant was collected after centrifugation of the cells and the secreted IL2 in the diluted cell culture medium (1:4 dilution) was measured using the IL2 Flex set cytometric bead array (CBA) (BD, 558270) following the manufacturer’s instructions. The acquisition was done on a BD Fortessa™ analyzer and the data was analyzed in FCAP Array™ v3.0.

PhosFlow cytometry analysis

Following stimulation, the cells were immediately fixed by adding the same volume of pre-warmed BD Cytofix Fixation Buffer (554655, BD) for 1 hr at 37 °C. After collecting the samples at all the different time points, they were then washed in FACS buffer and re-suspended in 200 μL of BD Phosflow Perm Buffer III (558050, BD) containing the antibodies for 30 min at 4 °C. After washing the cells with FACS buffer, they were re-suspended in FACS buffer to be acquired on the BD Fortessa™.

The antibodies used are the following (Table below): VIMP/SELS (V6639, Sigma-Aldrich ) (dilution 1:200) with Goat Anti-rabbit IgG H&L Alexa Fluor® 647 (A-21245, Invitrogen) (dilution 1:200), NFAT1 FITC (611060, BD) (dilution 1:50), phospho p38 MAPK (T180/Y182) Alexa Fluor 647 (562066, BD) (dilution 1:50), Anti-Human phospho NFATC1 (pS172) mAb (MAB5640, R&D Systems) (dilution 1:400), phospho NFATC2 (NFAT1) (S326) (SAB4503945, Sigma-Aldrich) (dilution 1:800), PE-Cy7 Mouse anti-ERK1/2 (pT202/pY204) (560116, BD) (dilution 1:50), phospho JNK1/2 (T182/Y185) (558268, BD), phospho cJun (S63) (9261S, Cell Signaling) (dilution 1:200), phospho p105 NFκB1 (S933) (4806S, Cell Signaling) (dilution 1:400), phospho p100 NFκB2 (S866/870) (4810S, Cell Signaling) (dilution 1:400), phospho p65 (S529) (558422, BD) (dilution 1:50), phospho RelB (S552) (4999S, Cell Signaling) (dilution 1:400) , Anti-Rabbit IgG H&L Alexa Fluor 647 (ab
150079, Abcam) (dilution 1:1000), APC Goat Anti-mouse IgG (minimal X-reactivity) (405308, Biolegend) (dilution 1:200). For the acquisition a BD Fortessa™ was used and the data was analyzed in FlowJo 7.6.5.

| Target                                      | Dilution | Company          | Clone (if applicable) | Reference |
|---------------------------------------------|----------|------------------|-----------------------|-----------|
| Anti-VIMP/SELS                              | 1:200    | Sigma-Aldrich    | Polyclone             | V6639     |
| FITC anti-NFATC2 (NFAT1)                    | 1:50     | BD Biosciences   | 1/NFAT-1              | 611960    |
| Mouse anti-human pNFATC1 (pS172) MAb       | 1:400    | R&D Systems      | 679340                | MAB5640   |
| APC Goat Anti-mouse IgG (minimal X-reactivity) Antibody | 1:200     | Biolegend        | N.A.                  | 405308    |
| Alexa Fluor 647 Mouse anti-NFkB p65 (pS529) | 1:50     | BD Biosciences   | K10-895.12.50         | 558422    |
| PE-Cy7 Mouse anti-ERK1/2 (pT202/pY204)      | 1:50     | BD Biosciences   | 20A                   | 560116    |
| Alexa Fluor 647 Mouse Anti-p38 MAPK (pT180/pY182) | 1:50      | BD Biosciences   | 36/p38                | 562066    |
| phospho NFAT1/NFATC2 (S326)                 | 1:800    | Sigma-Aldrich    | Polyclone             | SAB4503945|
| phospho c-Jun (S63)                         | 1:200    | Cell Signaling   | Polyclone             | 9261S     |
| phospho JNK1/2 (T183/Y185)                  | 1:200    | BD Biosciences   | Polyclone             | 558268    |
| phospho p105 NFkB1 (S933)                   | 1:400    | Cell Signaling   | 18E6                  | 4806S     |
| phospho p100 NFkB2 (S866/870)               | 1:400    | Cell Signaling   | Polyclone             | 4810S     |
| phospho RelB (S552)                         | 1:400    | Cell Signaling   | Polyclone             | 4999S     |
| Goat Anti-rabbit IgG H&L (Alexa Fluor® 647) | 1:200    | Invitrogen       | N.A.                  | A-21245   |

### Calcium/Ca2+ flux

To measure the calcium flux in Teffs, the cells were stained with mouse monoclonal [RPA-T4] anti-human CD4 FITC (555346, BD Biosciences) (dilution 1:100), LIVE/DEAD® Fixable Near-IR Dead Cell Stain (L10119, Thermo Fisher Scientific) (dilution 1:500) and the calcium dye Indo-1 (I203, Thermo Fisher Scientific) (5 uM) for 60 min at 37 °C in complete IMDM medium as for the culture of Teffs. Following 3 washes with medium the cells were re-suspended in 300uH of medium and incubated for another 15-30 min at 37°C. The baseline of the calcium signal was measured for approximately 30 sec before adding the soluble CD3/CD28 antibodies (1:40) (10971, StemCell) or 100 ng/ml Ionomycin (I0634, Sigma-Aldrich) to measure the activation-induced calcium flux. The cells were acquired on a BD Fortessa™ analyzer and the data was analyzed in FlowJo v10.5.

### Ethics statement
The study procedures were approved by the ethic committee of the Red Cross Luxembourg. Informed consent was obtained from healthy blood donors through the Red Cross Luxembourg.

**Statistical analysis**

P values were calculated with paired two-tailed Student t test (Graphpad Prism or Excel) as specified in Figure legend. If the other test was used, it has also been specified in the corresponding Figure legend. All error bars represent the standard deviation.

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