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# Adaptively introgressed Neandertal haplotype at the OAS locus functionally impacts innate immune responses in humans

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## Abstract

**Background:** The 2'-5' oligoadenylate synthetase (OAS) locus encodes for three OAS enzymes (OAS1-3) involved in innate immune response. This region harbors high amounts of Neandertal ancestry in non-African populations; yet, strong evidence of positive selection in the OAS region is still lacking.

**Results:** Here we used a broad array of selection tests in concert with neutral coalescent simulations to demonstrate a signal of adaptive introgression at the OAS locus. Furthermore, we characterized the functional consequences of the Neandertal haplotype in the transcriptional regulation of OAS genes at baseline and infected conditions. We found that cells from people with the Neandertal-like haplotype express lower levels of OAS3 upon infection, as well as distinct isoforms of OAS1 and OAS2.

**Conclusions:** We present evidence that a Neandertal haplotype at the OAS locus was subjected to positive selection in the human population. This haplotype is significantly associated with functional consequences at the level of transcriptional regulation of innate immune responses. Notably, we suggest that the Neandertal-introgressed haplotype likely reintroduced an ancestral splice variant of OAS1 encoding a more active protein, suggesting that adaptive introgression occurred as a means to resurrect adaptive variation that had been lost outside Africa.

**Keywords:** Natural selection, Introgression, Innate immunity

## Background

Whole-genome sequencing of several archaic human genomes [1–5] representing Neandertals and an as yet geographically and paleontologically unknown population referred to as Denisovans has revealed gene flow between these populations and the ancestors of present-day humans. Neandertal ancestry makes up approximately 0.5–2% of the ancestry of most living humans, with higher amounts of Neandertal ancestry found outside of Africa

[6–8]. While it seems that there may have been widespread purifying selection against Neandertal ancestry in humans [6, 8, 9], some positive selection on Neandertal genes (adaptive introgression) has also been observed [10, 11]. Neandertals and other archaic populations inhabited Eurasia for several hundred thousand years [12] and were likely well adapted to their environments. Therefore, some genetic variation inherited from these archaic humans may have been adaptive in modern humans, particularly across phenotypes that are strongly influenced by direct interactions with the surrounding environment [10], such as our immune response to infectious agents [13].

The oligoadenylate synthetase (OAS) locus on chromosome 12, which harbors three genes (OAS1, OAS2, OAS3) encoding the 2'-5' OAS enzymes has received considerable attention due to its clear signatures of multiple archaic

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our tests do not allow us to determine the exact timing when selection started to act on the Neanderthal alleles.

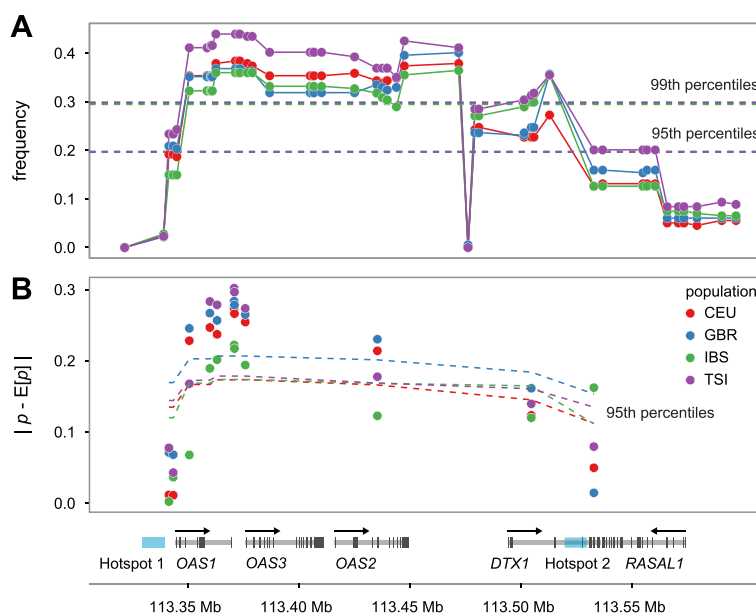
### Results

Mendez et al. [15] previously reported evidence of archaic introgression at the OAS locus. We first verified these results, using data from phase 3 of the 1000 Genomes Project [20], by examining the relationships of all modern human sequences at the OAS locus with the Altai Neanderthal, Denisovan, and an inferred ancestral sequence. Clustering the human haplotypes resulted in 10 consensus sequences representing human haplotype clusters, which we combined with the archaic sequences in a neighbor-joining tree (Fig. 1a; see “Methods”). This tree confirms that the Denisovan haplotype is not present in the population samples represented in the 1000 Genomes Project dataset. In contrast, the Neanderthal haplotype is found at relatively high frequencies outside of Africa, reaching highest frequencies in European population samples (up to 43 uc, Fig. 1b). Additionally, we corroborate the finding of Mendez et al. [15] that Neanderthal-like haplotypes in the OAS region are too long to have resulted from incomplete lineage sorting (ILS) ( $P \leq 2 \times 10^{-3}$ ; see “Methods”).

To investigate the hypothesis of non-neutral evolution at the OAS locus, we first tested whether the observed frequencies of Neanderthal-like sites (NLS) in the OAS region

are higher than expected under neutrality. We defined NLS as bi-allelic single nucleotide polymorphisms (SNPs) with derived alleles that are shared between Neandertals and a non-African population sample, but absent in a sub-Saharan African sample [6, 21, 22]. Specifically, we simulated the expected allele frequency of NLS under neutrality, using a demographic model based on previously inferred parameters of human demographic history [23–25] (Additional file 1: Figure S1, Additional file 2: Table S1). We considered both a model with a single pulse of Neanderthal introgression occurring over a span of 500 years into the ancestral Eurasian population after their population split from Africa and two additional two-pulse models (Additional file 1: Figure S1, Additional file 2: Table S1). We found that in all European populations (with the exception of Finnish (FIN)), the highest frequency NLS fall in the extreme 1% of all simulations and are significantly elevated in frequency (all five European samples pass false discovery rate (FDR) < 0.05; see Additional file 2: Table S2), regardless of whether we assume a single-pulse introgression model (Fig. 2a) or two-pulse introgression models (Additional file 1: Figure S2), indicating that the high frequencies of NLS found in most European populations likely reflects a history of adaptation.

We next sought to examine the consistency of the simulation results described above using genetic data



**Fig. 2** OAS-introgressed haplotypes are found at higher frequencies in European populations than expected under neutrality. **a** Comparison of frequency (*y*-axis) of NLS in the OAS locus in the CEU, GBR, IBS, and TSI European population samples with respect to neutral expectations (*dashed lines*) based on coalescent simulations. **b** Absolute difference between observed and expected allele frequency in the same four present-day European samples (*y*-axis) based on ancient DNA data. *Dashed lines* represent the 95th percentile of the expected distribution based on similar deviations calculated on a dataset of approximately 1 million SNPs scattered around the genome and with comparable present-day frequencies to those found for NLS in the OAS region

from a dataset of 230 ancient Eurasian individuals [26]. Assuming neutrality, the expected frequency of an allele in contemporary European populations can be predicted as a linear combination of allele frequencies sampled from representative ancient populations that have contributed ancestry to present-day European populations in different proportions [26, 27] (see “Methods”). Using this approach, we calculated the expected allele frequency in four present-day European samples from the 1000 Genomes Project [20] at the 11 NLS falling within the bounds of the three OAS genes, based on the ancient allele frequencies estimated by Mathieson et al. To set up our null expectations we performed a similar analysis on a dataset of approximately 1 million SNPs scattered around the genome, generated by Mathieson et al. (by merging 213 ancient samples dated between 6500 and 300 BCE with sequencing data from four European samples from the 1000 Genomes Project). We found that the NLS at OAS are outliers in the genome with respect to deviations from ancient frequencies. More specifically, we found that the allele frequencies of 6/11 OAS SNPs tested in the *OAS1*–*OAS3* region (those 6 SNPs fall within the block of SNPs with NLS frequency greater than 99% of neutral simulations) have increased above the frequency predicted by ancient Eurasian samples by more than 20%, significantly more than what we observed for other SNPs genome-wide with comparable present-day frequencies (lowest  $P = 0.00476$ , Fig. 2b, Additional file 2: Table S3). Our findings at this single locus are consistent with results from the genome-wide selection scan performed by Mathieson et al. [26] where the OAS region also showed evidence of selection ( $P < 10^{-7}$ ; Additional file 2: Table S3), even if it did not reach genome-wide significance after multiple test correction.

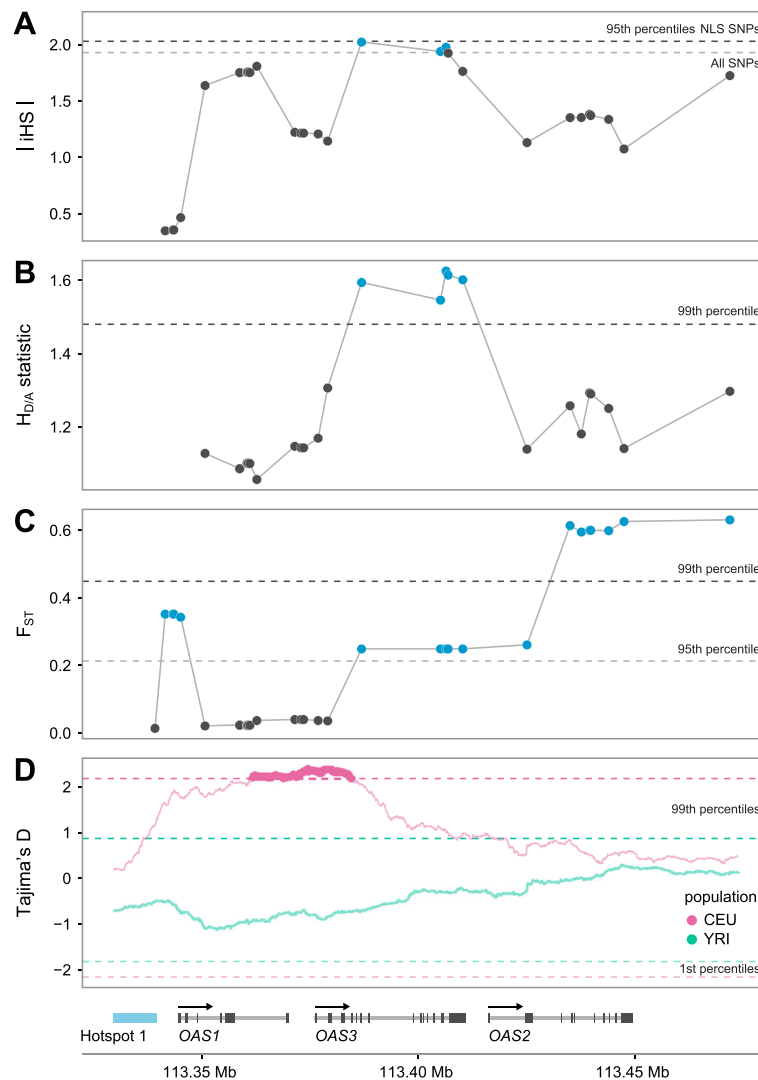
We next searched for additional evidence of recent selection, as measured by the iHS [28] and DIND [29] statistics. These tests share a similar rationale: an allele that has recently been driven to high population frequency by positive selection should be associated with unusually long-range LD (iHS) and reduced intra-allelic nucleotide diversity (DIND) [28–30]. We found that several NLS in the OAS region show significantly high iHS and DIND values with respect to genome-wide expectations (Fig. 3a and Additional file 1: Figure S3), further supporting that they have been targeted by positive selection.

As an additional means of understanding whether Neandertal haplotypes at OAS are longer than would be expected if they had evolved neutrally, we followed up these empirical haplotype-based tests with a simulation approach using a simple test statistic,  $H_{D/A}$ . This statistic (fully defined in “Methods”) compares pairwise haplotype homozygosity lengths within the set of haplotypes carrying a derived (Neandertal) allele and within those

carrying the ancestral allele. We utilized the same demographic models described above for our frequency-based test to understand whether the haplotypes surrounding derived alleles at NLS are longer than expected under a neutral model, conditional on the map of recombination in the OAS region. Again, we found that several derived Neandertal-like alleles are associated with a haplotype that is significantly longer than observed in simulated loci and that these results were robust to a series of alternative simulation models tested, including the one-pulse and two-pulse models described above and two additional one-pulse models which varied the mutation and recombination rates (Fig. 3b, Additional file 2: Table S4). A caveat to this simulation-based approach is that the parameters of the true demographic history of Neandertal-modern human admixture remain uncertain. Therefore, comparisons of neutral demographic models of Neandertal introgression to tests of selection utilizing haplotype lengths, such as  $H_{D/A}$ , are limited by the current state of demographic parameter estimates.

Finally, we calculated the levels of population differentiation between European and East Asian populations for all NLS in the OAS region. Interestingly, we found extreme levels of differentiation for most NLS ( $F_{st}$  as high as 0.6,  $P_{\text{empirical}} < 0.01$ , Fig. 3c. Additional file 2: Table S5), except within the genomic region covering *OAS1* and part of *OAS3*. These results suggest that NLS surrounding *OAS1* have been positively selected in both European and East Asian populations but that distinct haplotypes, possibly recombinant forms of an original introgressed Neandertal haplotype, have been selected in Europe and Asia.

Our population genetic results provide evidence that Neandertal alleles at the OAS locus have likely experienced positive selection during one or several phases after their introduction into the human population, suggesting a possible functional role of these alleles in human innate immune responses. To study this possibility, we analyzed RNA-sequencing (RNA-seq) data collected on primary macrophages from 96 European-descent individuals, before and after in vitro infection with *Salmonella typhimurium*. After 2 h of infection, we found that all OAS genes were strongly upregulated (up to 19-fold,  $P < 1 \times 10^{-10}$ ; Additional file 1: Figure S4), confirming the ability of *Salmonella* to activate the interferon (IFN) production pathway [31–33]. Using genotype data available for the same individuals (673 SNPs spanning the OAS region; see “Methods”), we tested if NLS were associated with variation in the expression levels of *OAS1*, *OAS2*, or *OAS3*, in either infected or non-infected macrophages. We found that among NLS, individuals that are heterozygous or homozygous for the Neandertal allele show reduced expression levels of *OAS3* (i.e. they were expression quantitative trait loci, or

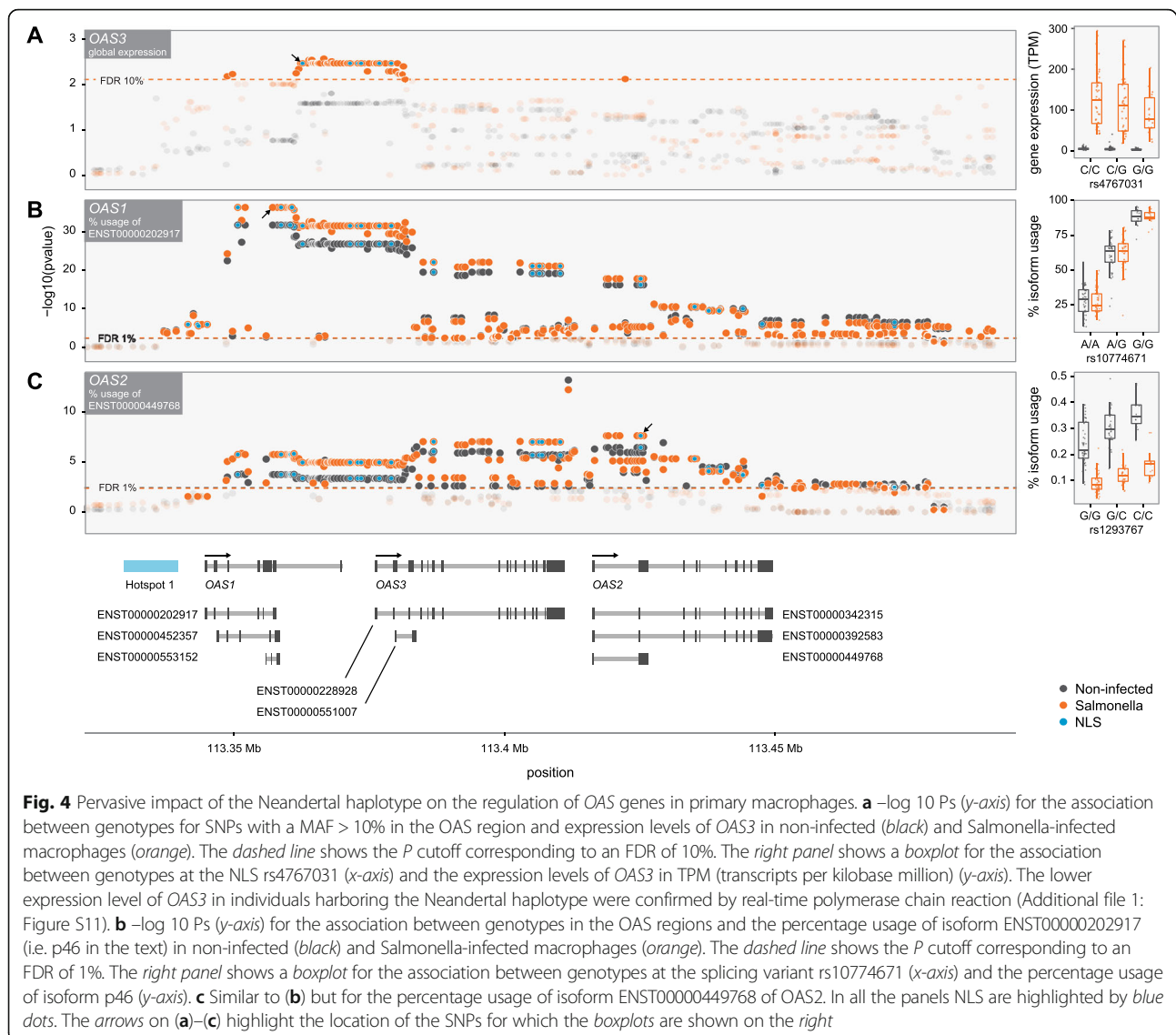


**Fig. 3** OAS-introgressed haplotypes show multiple signatures of positive selection. **a** Normalized absolute iHS scores (*y*-axis) across SNPs in the CEU sample. Dashed lines represent the 95th percentile of iHS calculated genome-wide across all SNPs (gray) and all NLS SNPs (black). **b** The  $H_{D/A}$  statistic in CEU (*y*-axis). Dashed line indicates the 99th percentile of  $H_{D/A}$  calculated across 1000 simulations. **c**  $F_{ST}$  calculated between CEU and all East Asian populations from the 1000 Genomes projects (*y*-axis). Dashed lines indicate the 95th (gray) and 99th (black) genome-wide percentiles of  $F_{ST}$ . Similar results are obtained when comparing any other European population against East Asians (Additional file 2: Table S5). **d** Tajima's D (*y*-axis) calculated for CEU and YRI samples. Dashed lines indicate the 1st and 99th genome-wide percentiles of Tajima's D

*cis* eQTL for *OAS3*) (FDR < 10%; Fig. 4a, Additional file 2: Table S6). Interestingly, these *cis* eQTL showed a much stronger effect in infected macrophages (best  $P_{salmonella} = 3.5 \times 10^{-3}$  versus best  $P_{non-infected} = 0.027$ ), supporting an interaction between the Neandertal haplotype and the *OAS3* response to *Salmonella* infection.

In addition to overall changes in expression, we took advantage of the power of RNA-seq data to test if NLS in the OAS regions influenced the ratio of alternative isoforms used for each of the OAS genes (i.e. alternative splicing QTL: asQTL). We found that SNPs associated with the Neandertal haplotype are significant asQTL for *OAS1* and *OAS2* in both infected and non-infected

macrophages (FDR < 1%; Fig. 4b, c). The effect of the splice site variant rs10774671 at determining what isoform is primarily encoded by *OAS1* was particularly strong ( $P \leq 2 \times 10^{-32}$ ). This SNP is also a strong asQTL and protein QTL in lymphoblastoid cell lines [34, 35]. The ancestral G allele at this SNP (AG at acceptor site) retains the splice site whereas the derived allele, A (AA at acceptor site) disrupts the splice site leading to the usage of a distinct isoform (Additional file 1: Figure S5). The Neandertal haplotype harbors the ancestral allele (encoding the p46 isoform), which is associated with high enzyme activity [36]. Interestingly, despite the fact that this ancestral allele is found at ~60–70% in most

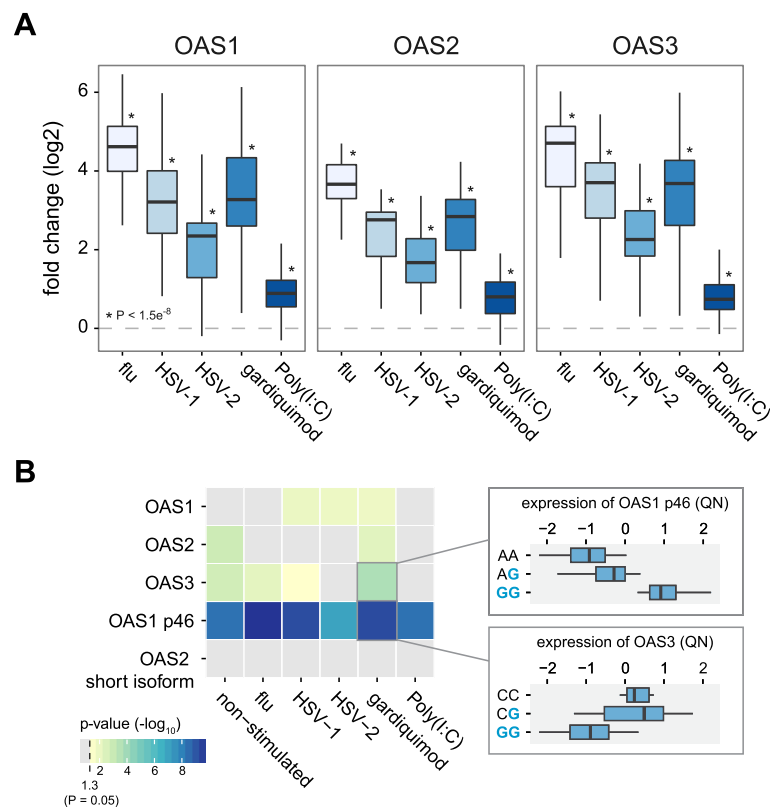


sub-Saharan African populations, outside of Africa this allele is only found on individuals with the Neandertal haplotype (with rare exceptions; ~2% of all haplotypes; Additional file 1: Figure S6), suggesting that the Neandertal segment is effectively reintroducing an ancestral variant that was already present in other human groups. To validate that this SNP had the same impact at determining what *OAS1* isoforms are expressed in African-descent individuals, we analyzed additional RNA-seq data collected from 41 African-American individuals. As among Europeans, rs10774671 is a strong asQTL for *OAS1* in both non-infected and infected macrophages ( $P < 1 \times 10^{-15}$ ; Additional file 1: Figure S7), despite laying in a non-Neandertal haplotype.

Because *OAS* genes are primarily involved in the control of viral infections, we decided to validate our functional findings on PBMCs from 40 individuals stimulated/

infected with viral-ligands (polyI:C and gardiquimod) and live viruses (Influenza, Herpes simplex virus [HSV] 1, and HSV2). The individuals were chosen based on their genotype for the NLS rs1557866, a SNP that is a strong proxy for the presence or absence of the Neandertal haplotype (9 were homozygous for the Neandertal haplotype, 15 were heterozygous, and 16 homozygous for the modern human sequence).

As expected, we found that all viral-associated immune triggers led to a marked increase in *OAS1-3* gene expression levels, as measured by real-time polymerase chain reaction (PCR) (up to 27-fold,  $P \leq 1.2 \times 10^{-8}$ , Fig. 5a), concomitantly with the upregulation of type-I and type-II interferon genes (Additional file 1: Figure S8). Confirming the QTL results obtained in macrophages, we found that rs10774671 was a strong asQTL for *OAS1* in both non-infected and infected PBMCs ( $P \leq 4.9 \times 10^{-5}$ , Fig. 5b).



**Fig. 5** The Neandertal haplotype in the OAS regions has a different impact on the regulation of OAS genes depending on the viral agents PBMCs are exposed to. **a** Log 2 fold induction (y-axis) of OAS1, OAS2, and OAS3 in response to different viral agents or viral-associated immune stimuli, relative to non-infected PBMCs. **b**  $-\log_{10} P$  for the association between genotype status for the Neandertal haplotype and overall expression levels of OAS genes and the expression of specific isoforms of OAS1 and OAS2 (those identified in Fig. 3 as associated with NLS). *Boxplots* show the association between genotype status (*blue* referring to Neandertal alleles) and expression levels of the p46 isoform of OAS1 and OAS3 in Gardiquimod-stimulated PBMCs

Likewise, we found that the presence of the Neandertal haplotype was associated with reduced expression levels of OAS3, particularly in PBMCs infected with influenza ( $P = 6.1 \times 10^{-3}$ ) and the synthetic ligand gardiquimod ( $P = 2.0 \times 10^{-4}$ ), which mimics a single strand RNA infection (Fig. 5b). Interestingly, the Neandertal haplotype harbors additional regulatory variants that only impact expression levels in a cell-type and immune stimuli specific fashion. For example, we found that the Neandertal haplotype is associated with increased expression levels of OAS2 in non-infected ( $P = 1.2 \times 10^{-3}$ ) and Gardiquimod-stimulated PBMCs ( $P = 4.9 \times 10^{-3}$ ), but neither in macrophages nor in PBMCs treated with other viral agents. Collectively, our functional data provide evidence for a pervasive impact of the Neandertal haplotype on the regulation of OAS genes that varies depending on the cell type and the immune stimuli to which the cells are responding.

## Discussion

The Neandertal lineage was present in Eurasia for at least 400,000 years [37], providing ample time for

Neandertals to adapt to local disease environments. The admixture process, which likely fostered the transmission of pathogens between Neandertals and humans migrating out of Africa, could have also led to the exchange of genes useful in responding to local pathogens. Here, we have demonstrated that a previously reported case of Neandertal introgression at the OAS locus [15] displays population genetic signatures that are characteristic of positive selection in the European population. We note, however, that our results are limited by an incomplete knowledge of the demographic history of admixture between Neandertals and modern humans and the fact that the neutrality tests we used are not perfectly suited to detect adaptive introgression. Yet, we have strengthened the case for adaptive introgression by providing direct functional evidence of a role for the Neandertal OAS haplotype in the regulatory responses in innate immune cells to infectious agents.

Our results show that the Neandertal haplotype at OAS is associated with several regulatory variants that reduce expression of OAS3 in response to infection, as

well as encode alternate isoforms of *OAS1* and *OAS2*. These dramatic functional implications of the Neandertal OAS haplotype support our case for adaptive introgression at OAS. Yet, because distinct functional polymorphisms segregate together in the same haplotype, inferring the exact variant(s) targeted by positive selection remains a daunting task. We speculate, however, that one of the strongest direct targets of selection is likely to have been the splice variant identified in *OAS1*.

The Neandertal haplotype carries the ancestral allele (G) of the *OAS1* splice variant (rs10774671), which is common both inside and outside of Africa. However, outside of Africa, the only haplotypes carrying this ancestral splice site are closely related to the Neandertal haplotype, with a few exceptions being rare recombined haplotypes (~2% of all haplotypes with the ancestral allele). This pattern reflects the possibility that Neandertal introgression, in effect, served as a means to resurrect the ancestral splice site from local extinction outside of Africa, probably following the out-of-Africa exodus. Interestingly, the same ancestral splice site is found in the Denisovan genome and may similarly have impacted adaptive introgression into the ancestors of Melanesians. While we cannot at present confirm that the ancestral splice site was missing from the ancestral Eurasian population, the presence of this allele only on the Neandertal haplotype hints at the possibility that this splice site was lost to drift and subsequently re-introduced by Neandertals, providing beneficial genetic variation at the OAS locus (Additional file 1: Figure S6). The Neandertal-introgressed allele encodes a protein variant (p46) that is associated with higher enzymatic activity [36]. The adaptive potential of this variant is supported by the observation that this variant (or other variants in strong LD with it) was shown to be associated with: (1) reduced infection and replication rates of West Nile virus ([38], but see [39]); (2) improved resistance to hepatitis C virus (HCV) infection [40, 41]; and (3) variable symptomology of tick-borne encephalitis (TBE) virus-induced disease (homozygous individuals for the Neandertal haplotype show the most severe symptoms of TBE). Strikingly, West Nile, hepatitis C, and TBE are all members of the *Flaviviridae* family, suggesting that Flaviviruses might have been the main drivers of selection in *OAS1*.

The differential responses of homozygous carriers of the Neandertal OAS haplotype to the different viruses described above suggest that the Neandertal haplotype is not uniformly beneficial in humans. Thus, it is plausible that both spatial and temporal fluctuations in virus populations could have led to fluctuating selection pressure on the Neandertal OAS haplotype, consistent with findings that genes in the immune system have been disproportionately targeted by positive selection since the dawn of agriculture [42].

Alternatively, and not mutually exclusively, alleles at OAS, and particularly the *OAS1* splice variant, might be evolving under balancing selection. This hypothesis is supported by the observation that the *OAS1* splice variant (rs10774671) is found at high frequency worldwide (0.11–0.7) and the significantly higher Tajima's D values observed around *OAS1*, as compared to genome-wide expectations (Fig. 3d). Moreover, *OAS1* is among the most diverse genes in both humans and non-human primates. Indeed, a recent analysis of genome-wide sequence data from a total of 55 individuals from four non-human ape species, chimpanzee (*Pan troglodytes ellioti*), bonobo (*Pan paniscus*), gorilla (*Gorilla gorilla gorilla*), and orangutan (*Pongo abelii*), identified *OAS1* as in the top 1% of genes showing the largest levels of nucleotide diversity among ape species [43], consistent with a scenario of long-term balancing selection (*OAS2* and *OAS3* are ranked in the 60th and 36th percentile of the genome-wide distribution, respectively) or, as previous research has suggested, rapid evolution across the primate order [44]. Further supporting the idea of balancing selection on the introgressed haplotype, our functional data suggest that the Neandertal haplotype contributes a range of gene expression responses in a cell-type and stimulus-specific manner.

Estimates of historical allele frequencies at the OAS locus, from ancient DNA samples dated between roughly 8500 and 2500 years ago, support the notion that the Neandertal haplotype did not follow a classic selective sweep model with constant directional selection. Under such a model, the observed present-day frequency of the Neandertal haplotype at the OAS locus of roughly 38% would suggest a codominant fitness effect of  $s \sim 0.0014$ – $0.0017$ , when assuming an initial frequency of 0.02 at introgression approximately 2000–2400 generations ago (see “Methods”). However, the observed allele frequency shift of 0.26 over the past 200–340 generations (maximum shift in CEU from ancient samples; see Fig. 2b) suggests that the selection coefficient associated with the Neandertal haplotype during this recent human evolution would have been  $s \sim 0.0044$ – $0.0075$ : 2.6–5.4 times larger than the above estimate. Therefore, both temporally varying selection and balancing selection could explain why the Neandertal haplotype at OAS did not show clear signatures of adaptive introgression in previous studies [15, 18].

Our study illustrates the difficulty of identifying strong candidates for adaptive introgression. In a genome-wide statistical framework, our use of neutral coalescent simulations as a null distribution for adaptive introgression likely would not have provided significant results after multiple test correction across loci. This suggests that other instances of less obvious adaptive introgression, especially those that did not follow a classic selective sweep model, may remain to be identified. Moving



forward, novel methods must be developed to identify such cases. Some progress is now being made on this front. For example, Racimo et al. [11] recently developed a genome-wide statistical framework which relies on distributions of uniquely shared derived alleles between humans and Neandertals (as does our study) to identify candidate regions for adaptive introgression. This study also highlighted the OAS locus as a candidate for adaptive introgression. Additionally, estimates of historical allele frequencies with increased spatial and temporal resolution provided by the sequencing of ancient human genomes are likely to play an important role in illuminating candidates for adaptive introgression which do not conform to classical selective sweep models.

## Conclusions

In conclusion, our study demonstrates that the frequency and haplotype distribution of Neandertal-like sites can be used in a neutral simulation framework that accounts for local genomic context to investigate the history of selection at a candidate locus for which genome-wide tests of selection provide ambiguous results. When combined with functional data, our results provide the strongest evidence to date in support of adaptive introgression in the OAS region. More generally, our study raises the possibility that adaptive introgression might not necessarily occur to select newly introduced variants but rather as a means to resurrect adaptive variation in modern human populations that had been lost due to demographic events.

## Methods

### Genome alignments and identification of Neandertal-like sites

Human/chimpanzee ancestral states were computed by parsimony using alignments from the UCSC Genome Browser for the human reference (hg19) and three outgroups chimpanzee (panTro2), orangutan (ponAbe2), and rhesus macaque (rheMac2) [45]. Ancestral state was assumed to be the chimpanzee allele (if available) if its state was confirmed by matching either orangutan or macaque. All sites with no inferred ancestral state were removed from our analysis.

We filtered the Altai Neandertal genome [4] and Denisovan genome [3] using the map35\_50 set of minimum filters provided at ([https://bioinf.eva.mpg.de/altai\\_minimal\\_filters/](https://bioinf.eva.mpg.de/altai_minimal_filters/)). For the frequency analysis, we combined these filtered datasets with 10 samples from outside of Africa (5-Eur European: CEU, FIN, GBR, IBS, TSI; 5-East Asian: CDX, CHB, CHS, JPT, KHB; see Additional file 2: Table S7 for population codes) and one sub-Saharan African sample YRI (Yoruba in Ibadan, Nigeria) from the 1000 Genomes Project Phase 3 [20], which we downloaded from (<https://mathgen.stats.ox.ac.uk/impute/1000GP%20Phase%203%20haplotypes%206%20October%202014.html>).

We first extracted all bi-allelic variants in our human, Neandertal, and chimpanzee alignments. We considered as NLS only those variants where the African sample (YRI) had a derived allele frequency of zero and both the non-African sample and Neandertal carried the derived allele. We restricted the follow-up haplotype-based analysis to only the CEU sample. In that analysis, we required the derived allele to be present in CEU in at least two copies in order to calculate our haplotype-based test statistic ( $H_{D/A}$ ).

### Haplotype clustering in the 1000 Genomes Project data

To assess the relationships and frequencies of Neandertal-like haplotypes in the OAS region, we identified haplotype clusters based on sequence similarity. First, we filtered all 5008 haplotypes in the 1000 Genomes Project phase 3 dataset to include only SNPs within the bounds of the three OAS genes (hg19 chr12: 113344739–113449528) and at which the Altai Neandertal [4] and Denisovan [3] genomes carry homozygous genotypes. Next, we clustered the human haplotypes such that all haplotypes in a cluster had no more than 60 pairwise differences, which resulted in 10 clusters. We inferred a consensus haplotype for each cluster based on the majority allele at each position in the cluster. Finally, we produced a neighbor-joining tree of the human cluster consensus sequences and the Altai, Denisovan, and ancestral sequences using the R package “ape.” We estimated confidence values for nodes with 1000 bootstraps.

We then calculated haplotype frequencies for each of the 1000 Genomes Project population samples based on the assignment of each haplotype to the clusters described above.

### Demographic model and neutral coalescent simulations

We performed coalescent simulations of the demographic history of the European, East Asian, African, and Neandertal populations applied by Vernot and Akey [25] based on previously inferred demographic models [23, 24]. In our first set of simulations, which were performed with ms [46], we extracted only allele frequencies at NLS. We gathered a total of 1 million simulation results. In each simulation run, a result was collected if a NLS was present in either the European or the East Asian population sample (or both). Therefore, null distributions specific to Europeans or East Asians included less than 1 million results (but typically > 800,000). Haplotype simulations were performed with macs [47] in order to explicitly simulate the genetic map (downloaded with the 1000 Genomes samples at the link above) of the 600 kb region centered on OAS (chr12:113100000–113700000). The general shape of the demographic model is illustrated in Additional file 1: Figure S1. Our baseline simulations were performed with the parameters specified in Additional file 2: Table S1,

assuming 25 years per generation and a mutation rate of  $2.5 \times 10^{-8}$  per bp per generation. Additional file 2: Table S1 also provides all parameters used in one-introgression and two-introgression pulse simulations. Additionally, we examined the one-pulse model with a slower mutation rate ( $1.25 \times 10^{-8}$ ) and a one-pulse model with a uniform recombination rate ( $1.3 \times 10^{-8}$ ; a value conservatively lower than the average for the OAS region,  $1.7 \times 10^{-8}$ ). Sample ms and macs commands are given at the end of this section.

For haplotype-based simulations, data were thinned in a manner similar to Sankararaman et al. [6] to account for imperfect SNP ascertainment in the 1000 Genomes dataset, such that SNPs with minor allele count of 1, 2, 3, 4, 5, 6, 7, 8, 9, and  $\geq 10$  were accepted with probabilities 0.25, 0.5, 0.75, 0.8, 0.9, 0.95, 0.96, 0.97, 0.98, and 0.99, respectively. Additionally, we only kept SNPs that were polymorphic in the simulated CEU sample. Finally, we performed an additional thinning of SNPs with uniform probability of 0.05 of removal to account for slightly elevated SNP density in the simulated data. The resulting simulated datasets had an average SNP density of 3.8 SNPs per kb compared to 2.5 in the real data. This is a slightly larger than ideal difference in SNP density, but we note that neither derived allele frequency nor our primary haplotype-based test statistic (described below) should be particularly sensitive to SNP density. In fact, Additional file 1: Figure S9 illustrates that our statistic is conservative with respect to SNP density. Additionally, our results hold in a simulation with half the mutation rate of our baseline simulation. This simulation resulted in an average SNP density of 1.9 SNPs per kb. So our observation of long haplotypes surrounding NLS is robust to SNP density across our simulations.

Sample ms command:

```
ms 752 1 -s 1 -I 4 250 250 250 2 0 -n 1 58.0027359781
-n 2 70.0410396717 -n 3 187.549931601 -n 4 0.2051
98358413 -eg 0 1 482.67144247 -eg 0 2 505.592963281
-eg 0 3 720.224280773 -em 0 1 2 0.358619661426 -em 0
2 1 0.358619661426 -em 0 1 3 0.111889334365 -em 0 3
1 0.111889334365 -em 0 2 3 0.446122858814 -em 0 3 2
0.446122858814 -en 0.00699726402189 1 1.98002735978
-eg 0.00699726402189 2 0.0 -eg 0.00699726402189 3
17.5076517287 -en 0.03146374829 2 2.03666547538 -en
0.03146374829 3 0.700185007205 -ej 0.0641347992705 3
2 -em 0.0641347992705 1 2 0.00015 -em 0.064134799
2705 2 1 0.00015 -em 0.0839811779742 2 4 0.00075 -em
0.0846651725023 2 4 0 -ej 0.0957592339261 2 1 -en
0.202462380301 1 1.0 -ej 0.957592339261 4 1.
```

Sample macs command:

```
macs 416 600000 -t 0.000731 -R oas_recreates.txt -I 4
216 198 0 2 0 -n 1 58.0027359781 -n 2 70.0410396717
-n 3 187.549931601 -n 4 0.205198358413 -eg 0 1 482.67
144247 -eg 1e-12 2 460.409556336 -eg 2e-12 3 720.22
4280773 -em 3e-12 1 2 0.4441436762 -em 4e-12 2 1
```

```
0.4441436762 -em 5e-12 1 3 0.138572826974 -em 6e-12 3
1 0.138572826974 -em 7e-12 2 3 0.552514733193 -em 8e-
12 3 2 0.552514733193 -en 0.00699726402189 1 1.9
8002735978 -eg 0.00699727402189 2 0.0 -eg 0.006997
28402189 3 18.896348561 -en 0.03146374829 2 2.7939
9962717 -en 0.03146375829 3 0.670250141711 -ej 0.05178
5044418 3 2 -em 0.051785054418 1 2 0.00015 -em 0.05
1785064418 2 1 0.00015 -em 0.0555806720117 2 4 0.00075
-em 0.0562646665398 2 4 0 -ej 0.0957592339261 2 1 -en
0.202462380301 1 1.0 -ej 0.957592339261 4 1 -h 1e3.
```

### Frequency and haplotype-based tests of neutrality using simulations

We examined the consistency of genetic variation with our neutral model using several approaches. First, we examined the likelihood of observing (Neandertal) allele frequencies as high as the OAS locus. Under neutrality, allele frequency is not dependent upon recombination rate; therefore, we can estimate the likelihood of our observed NLS frequency in the OAS region. To create a distribution of expected NLS frequency in each introgression model, we first extracted derived allele counts (DAC) at simulated NLS for samples of 250 individuals each for Europeans and East Asians in the simulated model. Next, for each non-African population sample from the 1000 Genomes Project, we binomially sampled the DACs from the appropriate simulated population (European for CEU, GBR, IBS, TSI; East Asian for CDX, CHB, CHS, JPT, KHV) using the true population sample size to create a distribution of NLS frequencies that is specific to each population sample (Fig. 2a, Additional file 1: Figure S3). Minimum *P* values across the OAS locus for each population are listed in Additional file 2: Table S2.

Additionally, we wanted to examine if the haplotypes carrying NLS at the OAS locus are longer than expected under neutrality when conditioning on the observed frequencies and the underlying genetic map, which would provide an additional signature of selection on introgressed haplotypes beyond empirical haplotype signatures. For this purpose, we modified a simple haplotype statistic *H* [48], which measures the average length of pairwise homozygosity tracts in base pairs—a quantity that is very straightforward to interpret. As selective sweeps are expected to create long haplotypes around the selected site, the *H* statistic should be higher in samples containing positively selected haplotypes compared to samples containing neutrally evolving haplotypes, when frequency and recombination are properly controlled, similar to other statistics based on haplotype lengths, such as *EHH*, *iHS*, and *nSL* [28, 30, 49]. However, in contrast to these other statistics, *H* does not require specification of analysis parameters such as minimum haplotype homozygosity levels below which haplotypes are no longer extended.

Under adaptive introgression, we specifically expect the introgressed (derived allele carrying) haplotypes to be longer than the ancestral haplotypes, due to the fact that selected Neandertal haplotypes will, on average, spend less time at low frequency, where they have a greater opportunity for recombination with non-introgressed haplotypes than neutral Neandertal haplotypes. We therefore defined our test statistic,  $H_{D/A}$ , as:

$$H_{D/A} = \ln(H_D/H_A);$$

where  $H_D$  and  $H_A$  are  $H$  calculated across haplotypes carrying derived NLS allele versus the ancestral allele (see Additional file 1: Figure S10).

For each model we generated 1000 simulated OAS regions containing NLS that reach a frequency greater than the 99th percentile for CEU in our frequency simulations. We calculated  $H_{D/A}$  for every simulated NLS and for every NLS in our true sample. We then compared our simulation results to the observed data in two ways. First, to visually examine the distribution of true  $H_{D/A}$  values to neutral simulations, we took the mean  $H_{D/A}$  score across all NLS in each of 1000 simulations and calculated the 95th and 99th percentiles of those 1000 simulations (Fig. 3b). Second, to examine the probability under neutrality of the observed  $H_{D/A}$  values at the peak of  $H_{D/A}$  between chr12:113380000–113420000, we compared the mean  $H_{D/A}$  score for SNPs in this window to the mean of SNPs in this window for each simulation that produced a NLS in that window. These results for all tested models are listed in Additional file 2: Table S4.

### Analysis of ancient Eurasian data

We utilized supplementary data Table 3 from Mathieson et al. [26]. This table includes maximum likelihood allele frequency estimates for three ancient population samples (HG: hunter-gatherer, EF: early farmer, SA: Steppe ancestry) and four present-day European samples from the 1000 Genomes Project (see the “Genome-wide scan for selection” section of methods in [26]). We intersect this table with allele frequencies for 1000 Genomes Yorubans (YRI) and the Altai Neandertal genotypes and only analyze sites for which we have data for all samples (1,004,612 SNPs).

To calculate the expected allele frequency in modern samples under drift, we used the estimated proportions ( $m$ ) of (HG, EF, SA) in each of the four present-day samples estimated by Mathieson et al. [26]: CEU = (0.196, 0.257, 0.547), GBR = (0.362, 0.229, 0.409), IBS = (0, 0.686, 0.314), and TSI = (0, 0.645, 0.355). We calculated the expected frequency  $E[p]$  of site as:

$$E[p] = \sum_i^{\{HG,EF,SA\}} (p_i \times m_i)$$

Next, we calculated the absolute difference between observed and expected allele frequency in all four present-day European samples at all available sites. To test the null hypothesis that OAS NLS have not changed in frequency more than expected under neutrality at 11 SNPs in the central OAS region (chr12:113200000–113600000), we first calculated the fraction of all autosomal SNPs in the dataset at similar present-day frequency (within 1% in the folded frequency spectrum) of each OAS NLS. Finally, we calculated the fraction of autosomal SNPs with an absolute observed minus expected frequency difference greater than or equal to the OAS NLS. These results are given in Additional file 2: Table S3 and illustrated in Fig. 2b.

This test does not explicitly incorporate variance in estimated ancient allele frequency. However, any bias in ancient allele frequency estimation should be distributed randomly across the genome. Therefore, our comparison to a genome-wide distribution of SNPs at similar present-day frequency should incorporate most of this error. Nonetheless, the selection test performed by Mathieson et al. [26] does incorporate such error, so we can also look to the  $P_s$  from that test to ensure consistency with our results.

### Estimation of selection coefficients

To estimate the selection coefficient  $s$  under constant positive selection for a given starting frequency ( $x_0$ ), final frequency ( $x_1$ ), and number of generations between these estimates ( $\Delta t$ ), we assumed a model of standard logistic growth of a co-dominant allele:

$$x_1 = \frac{x_0}{x_0 + (1-x_0)e^{-s\Delta t}}$$

This equation can be easily solved to obtain  $s$ , given  $x_0$ ,  $x_1$ , and  $\Delta t$ .

### Probabilities that Neandertal haplotypes are shared due to incomplete lineage sorting

We estimated the probability that Neandertal haplotypes observed in the OAS region could be a product of ILS using the method outlined by Huerta-Sanchez et al. [50]. This method estimates the probability that a haplotype of length ( $H$ ) is shared due to incomplete lineage sorting as:

$$P(H) = 1 - \text{gammaCDF}(H, \text{shape} = 2, \text{rate} = 1/L);$$

where  $L$  is the expected shared length,  $L = 1/(r \times t)$  where  $r$  is recombination rate and  $t$  is time in generations.

We first estimated the maximum haplotype length observed in each of the 1000 Genomes Project population samples by identifying tracts of identity-by-state (IBS)

that include NLS between the Altai Neandertal haplotype (considering only homozygous positions) and individual human haplotypes. We identified all tracts of IBS bounded by NLS (using YRI as the African reference) and took the maximum of these per-population sample IBS distributions.

We identified the maximum tract length (198,441 bp) in European (and some admixed Native American) samples. Maximum tract length for East Asian samples (54,385 bp) is notably shorter. As expected, no IBS tracts were identified in African population samples.

Assuming a conservatively low recombination rate for the OAS region ( $1.3 \times 10^{-8}$ ) and a conservatively short divergence between Neandertals and humans of 300,000 years (and 25 years per generation), the probability of a length of at least 54,385 is  $P(54,385) = 0.002$ . The probability of the longest tracts is much smaller,  $P(198,441) = 1.15 \times 10^{-12}$ . Therefore, it is unlikely that Neandertal haplotypes are shared with non-Africans due to shared ancestral variation.

We repeated this analysis using the Denisovan genome and found no IBS tracts defined by Denisovan-like sites, suggesting that introgressed Denisovan haplotypes described by Mendez et al. [17] are not present in the 1000 Genomes Project Samples, consistent with our findings in Fig. 1a.

#### Neutrality tests in phase 3 of the 1000 Genomes data

We calculated *iHS*, *DIND*, *Fst*, and Tajima's *D* on different populations from phase 3 of the 1000 Genomes project. The phased data were downloaded from Impute reference data panel ([https://mathgen.stats.ox.ac.uk/impute/1000GP\\_Phase3.html](https://mathgen.stats.ox.ac.uk/impute/1000GP_Phase3.html)) and were filtered for bi-allelic SNPs. Arlequin version 3.5.1.3 [51] was then used to calculate *Fst* estimates derived from ANOVA among all pairs of European and Asian populations. *DIND* was calculated as previously described [29] and Tajima's *D* was calculated for all SNPs using a window of 25 kb either side of each SNP (i.e. in window of 50 kb in total). *iHS* values were recovered from a genomic scan performed using hapbin [52] on the same phase 3 release of the 1000 Genomes dataset.

#### Sample collection

Buffy coats from 96 healthy European American donors and 41 African Americans were obtained from Indiana Blood Center (Indianapolis, IN, USA). Only individuals self-reported as currently healthy and not under medication were included in the study. The individuals recruited in this study were men aged 18–55 years old. For 80% of our samples, we have information on their exact age (for the remaining 20% of donors we only know that they were adults aged less than 55 years) and we found no association between the presence of the Neandertal haplotype (as measure by rs1557866) and age ( $P > 0.5$ ). Thus, variation in age is not likely impact

any of our conclusions. In addition to self-identified ancestry labels, we used the genome-wide genotype data (see the “DNA extraction and genotyping” section) to estimate genome-wide levels of European and African ancestry in each sample using the program ADMIXTURE [53]. Consistent with previous reports, we found that self-identified European Americans showed limited levels of African admixture (mean = 0.4%, range 0–13%). The African Americans used in our study were selected on having at least 75% of African ancestry.

#### DNA extraction and genotyping

DNA from each of the blood donors was extracted using the Gentra Pure Gene blood kit (Qiagen). Genotyping of each individual was then performed by Illumina's HumanOmni5Exome bead Chip array and complemented with imputed data from the 1000 Genomes data using Impute2 [54]. Here, we only focused on genetic diversity surrounding the OAS region – chr12:113229549–113574044 (~344 kb) spanning from the beginning of *RPH3A* to the end of *RASAL1*—for a total of 673 SNPs with a MAF above 10%. Genotypes can be found in Additional file 2: Table S8.

#### Isolation of monocytes and differentiation of macrophages

Blood mononuclear cells were isolated by Ficoll-Paque centrifugation. Monocytes were purified from PBMCs by positive selection with magnetic CD14 MicroBeads (Miltenyi Biotech) using the autoMACS Pro Separator. All samples had purity levels above 90%, as measured by flow cytometry using an antibody against CD14 (BD Biosciences). Monocytes were then cultured for seven days in RPMI-1640 (Fisher) supplemented with 10% heat-inactivated FBS (FBS premium, US origin, Wisent), L-glutamine (Fisher), and M-CSF (20 ng/mL; R&D systems). Cell cultures were fed every two days with complete medium supplemented with the cytokines previously mentioned. Before infection, we systematically verified that the differentiated macrophages presented the expected phenotype for non-activated macrophages (CD1a+, CD14+, CD83–, and HLA-DRlow (BD Biosciences)).

#### Bacterial preparation and infection of macrophages

The day prior to infection, aliquots of *Salmonella typhimurium* (Keller strain) were thawed and bacteria were grown overnight in Tryptic Soy Broth (TSB) medium. Bacterial culture was diluted to mid-log phase prior to infection and supernatant density was checked at OD600. Monocyte-derived macrophages were then infected with *S. typhimurium* at a multiplicity of infection (MOI) of 10:1. A control group of non-infected macrophages was treated the same way but using medium without bacteria. After 2 h in contact with the bacteria, macrophages were

washed and cultured for another hour in the presence of 50 mg/mL of gentamycin in order to kill all extracellular bacteria present in the medium. The cells were then washed a second time and cultured in complete medium with 3 mg/mL gentamycin for an additional 2 h, the time point to which we refer in the main text.

#### Infection/stimulation of PBMC

PBMCs from a subset of 40 individuals used to derive macrophages were cultured in RPMI-1640 (Fisher) supplemented with 10% heat-inactivated FBS (FBS premium, US origin, Wisent) and 1% L-glutamine (Fisher). The 30 individuals were chosen based on their genotype for rs1557866, a SNP which derived allele is of Neanderthal origin and that we used as a proxy to identify individuals harboring the Neanderthal haplotype in the OAS region. From the 40 individuals, and based on this SNP, nine individuals were homozygous for the Neanderthal haplotype, 15 were heterozygous, and 16 homozygous for the modern human sequence.

For each of the tested individuals, PBMCs (1 million per condition) were stimulated/infected with one of the following viral-associated immune challenges: polyI:C (10 µg/mL, TLR3 agonist), Gardiquimod (0.5 µg/mL, TLR7 and TLR8 agonist), Influenza PR8 WT (multiplicity of infection (MOI) of 0.05:1), HSV 1 ( $1.55 \times 10^2$  CPE), and HSV2 ( $19.5 \times 10^4$  CPE). PBMCs were stimulated/infected for 4 h with TLR ligands and Influenza and 6 h with HSV1 and HSV2. A control group of non-infected PBMC was treated the same way but with only medium.

#### RNA extraction, RNA-seq library preparation, and sequencing

Total RNA was extracted from the non-infected and infected/stimulated cells using the miRNeasy kit (Qiagen). RNA quantity was evaluated spectrophotometrically and the quality was assessed with the Agilent 2100 Bioanalyzer (Agilent Technologies). Only samples with no evidence of RNA degradation (RNA integrity number > 8) were kept for further experiments. RNA-seq libraries were prepared using the Illumina TruSeq protocol. Once prepared, indexed complementary DNA (cDNA) libraries were pooled (6 libraries per pool) in equimolar amounts and sequenced with single-end 100 bp reads on an Illumina HiSeq2500. Results based on the entire dataset are described elsewhere [55]. Here, we only studied transcript-level and gene-level expression estimates for *OAS1*, *OAS2*, and *OAS3*. The transcript-level and gene-level expression data for these genes can be found in Additional file 2: Table S9.

#### Quantifying gene expression values from RNA-seq data

Adaptor sequences and low quality score bases (Phred score < 20) were first trimmed using Trim Galore (version

0.2.7). The resulting reads were then mapped to the human genome reference sequence (Ensembl GRCh37 release 65) using TopHat (version 2.0.6) and using a hg19 transcript annotation database downloaded from UCSC. Gene-level expression estimates were calculated using featureCounts (version 1.4.6-p3) and transcript-level expression values were obtained using RSEM under default parameters.

#### Quantitative real-time PCR

For the PBMC samples, we measured the expression levels of OAS and interferon genes using real-time PCR. A total of 100 ng of high-quality RNA was reverse-transcribed into cDNA using the qScript cDNA Super-Mix (Quanta Biosciences). Quantitative real-time PCR was performed using 96.96 Dynamic Array™ IFCs and the BioMark™ HD System from Fluidigm. For the TaqMan gene assays, we used the following TaqMan Gene Expression Assay (Applied BioSystems) to quantify the expression levels of interferon genes: *IFNA1* (Hs03044218), *IFNA6* (Hs00819627), and *IFNG* (Hs00989291). To quantify the overall expression levels of OAS genes, we used probes that capture all common isoforms of *OAS1* (Hs00973635), *OAS2* (Hs00942643), and *OAS3* (Hs00196324). Custom-made probes were designed to specifically target the short-isoform of *OAS2* (Forward Primer Sequence CTGCAGGAACCCGAACAGTT; Reverse Primer Sequence ACTCATGGCCTAGAGGTTGCA; Reporter Sequence AGAGAAAAGCCAAAGAA). As housekeeping genes, we used: *GAPDH* (Hs02758991), *GUSB* (Hs99999908), *HPRT1* (Hs99999909), and *POLR2A* (Hs00172187). The results reported in the manuscript used *POLR2A* as a reference but all conclusions remain unchanged when using any of the other housekeeping genes.

We start by doing a preamplification of the cDNA using the PreAmp Master Mix (Fluidigm). Preamplified cDNA was then diluted 2× on a solution of 10 mM Tris-HCl (pH 8.0) and 0.1 mM EDTA. In order to prepare samples for loading into the integrated fluid circuit (IFC), a mix was prepared consisting of 360 µL TaqMan Fast Advanced Master Mix (Applied BioSystems) and 36 µL 20× GE Sample Loading Reagent (Fluidigm). A total of 2.75 µL of this mix was dispensed to each well of a 96-well assay plate and mixed with 2.25 µL of preamplified cDNA. Following priming of the IFC in the IFC Controller HX, 5 µL of the mixture of cDNA and loading reagent were dispensed in each of the sample inlet of the 96.96 IFC. For the TaqMan gene assays, 5 µL of mixes consisting of 2.5 µL 20× TaqMan Gene Expression Assay (Applied BioSystems) and 2.5 µL 2× Assay Loading Reagent (Fluidigm) were dispensed to each detector inlet of the 96.96 IFC. After loading the assays and samples into the IFC in the IFC Controller HX, the

IFC was transferred to the BioMark HD and PCR was performed using the thermal protocol GE 96 × 96 Fast v1.pcl. This protocol consists of a Thermal Mix of 70 °C, 30 min; 25 °C, 10 min, Hot Start at 95 °C, 1 min, PCR Cycle of 35 cycles of (96 °C, 5 s; 60 °C, 20 s). Data were analyzed using Fluidigm Real-Time PCR Analysis software using the Linear (Derivative) Baseline Correction Method and the Auto (Detectors) Ct Threshold Method.

To quantify the expression levels of the *OAS1* isoform associated with the derived allele at the splicing variant rs10774671, we used SybrGreen and the following forward (GCTGAGGCCTGGCTGAATTA) and reverse (CCACTTGTTAGCTGATGTCCTTGA) primers. PCR was performed using the thermal protocol 50 °C, 2 min; 95 °C, 10 min, PCR cycle of 40 cycles of (95 °C, 15 s; 60 °C, 1 min). A melting curve was also performed to check for non-specific amplification.

### Genotype–phenotype association analysis

eQTL and asQTL were performed against *OAS1*, *OAS2*, and *OAS3*. We examined associations between SNP genotypes and the phenotype of interest using a linear regression model, in which phenotype was regressed against genotype. In particular, expression levels were considered as the phenotype when searching for eQTL and the percentage usage of each isoform in each gene when mapping asQTL. To avoid low power caused by rare variants, only SNPs in the *OAS* region with a minor allele frequency of 10% across all individuals were tested (i.e. 673 SNPs within the region chr12:113229549–113574044). In all cases, we assumed that alleles affected the phenotype in an additive manner. For the eQTL and asQTL analyses on macrophages, we mapped Salmonella-infected and non-infected samples separately. We controlled for FDRs using an approach analogous to that of Storey and Tibshirani [56], which makes no explicit distributional assumption for the null model but instead derives it empirically null from permutation tests, where expression levels were permuted 1000 times across individuals. For the non-infected and infected/stimulated PBMCs we only tested expression levels against the SNPs identified as eQTL or asQTL in the macrophage data (specifically, the SNPs for which boxplots are shown in Fig. 4).

### Additional files

**Additional file 1:** Supplementary figures. (PDF 2303 kb)

**Additional file 2:** Supplementary tables. (XLSX 1030 kb)

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### Availability of data and materials

The datasets supporting the conclusions of this article are included within the article and in Additional file 2: Tables S8 and S9. The raw data were deposited under GEO accession numbers GSE73765 and GSE81046.

### Authors' contributions

Conception and design: AJS, PWM, LBB; acquisition of data: AJS, AD, YN, VY, LBB; analysis and interpretation of data: AJS, AD, YN, PWM, LBB; contributed unpublished, essential data, or reagents: CA, JET; drafting or revising the article: AJS, PWM, LBB. All authors read and approved the final manuscript.

### Competing interests

The authors declare that they have no competing interests.

### Ethics approval and consent to participate

Institutional ethics (IRB) approval was obtained by the CHU Sainte-Justine Institutional Review Board (approved project #3557) and all subjects gave written informed consent prior to participation. Experimental methods comply with the Declaration of Helsinki.

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