Title: Adaptively introgressed Neandertal haplotype at the OAS locus functionally
impacts innate immune responses in humans.

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Abstract

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. Here we used a broad array of

selection tests in concert with neutral coalescent simulations to firmly 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*. Notably, the

Neandertal-introgressed haplotype 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 percent 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 OAS locus on chromosome 12, which harbors three genes (*OAS1, OAS2,*

OAS3) encoding the 2'-5' oligoadenylate synthetase enzymes has received

considerable attention due to its clear signatures of multiple archaic haplotypes in

populations outside of Africa [14,15], and the critical role of OAS genes in the

innate immune response to viruses [16]. Mendez and colleagues [17] first

identified an introgressed haplotype of *OAS1* from Denisovans that is restricted to

individuals in Indonesia and Melanesia. Later, these authors [15] identified a

Neandertal haplotype at the OAS locus that spans a ~190 kilobase region between

two surrounding recombination hotspots.

The elevated frequency of the Neandertal derived alleles in the OAS locus (Figure

1) relative to average levels of Neandertal ancestry in Europeans [6] , along with

the key role OAS genes play in protective immunity against viral infections raises

the possibility that introgressed Neandertal haplotypes at OAS may have been

adaptive in modern humans. While some studies provide suggestive evidence of

adaptive introgression at the OAS locus [6,11], strong evidence of positive

selection in the OAS region is still lacking. Indeed, several studies failed to reject

a model of neutral evolution for the Neandertal haplotype when using standard

neutrality tests [15,18].

We hypothesize that the overall lack of signals of selection in the OAS region

stems from the low power of standard neutrality tests to detect adaptive

introgression [19]. Here we circumvent this issue by testing the hypothesis of

adaptive introgression using extensive neutral coalescent simulations specifically

tailored to match the genomic features of the OAS region in combination with

several empirical observations of Eurasian genetic variation, and ancient DNA

data from Eurasia. We firmly demonstrate a population genetic signal of adaptive

introgression at the OAS locus and characterize the functional consequences of the

Neandertal haplotype in the transcriptional regulation of OAS genes in

macrophages and peripheral blood mononuclear cells (PBMCs) at baseline and

infected conditions. We note that the term 'adaptive introgression' will be used in

a broad sense, as 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 Neandertal, 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 (Methods, Figure 1A, S1). This tree

confirms that the Denisovan haplotype is not present in the population samples

represented in the 1000 Genomes Project dataset. In contrast, the Neandertal

haplotype is found at relatively high frequencies outside of Africa, reaching

highest frequencies in European population samples (up to 43%, Figure 1B).

Additionally, we corroborate the finding of Mendez and colleagues [15] that

Neandertal-like haplotypes in the OAS region are too long to have resulted from

incomplete lineage sorting (ILS) ($P \leq 2 \times 10^{-3}$, Methods).

To investigate the hypothesis of non-neutral evolution at the OAS locus we first

tested whether the observed frequencies of Neandertal-like sites (NLS) in the OAS

region are higher than expected under neutrality. We defined NLS as bi-allelic

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](Figure S1, Table S1). We considered both a model

with a single pulse of Neandertal 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 (Figure S1, 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 FDR < 0.05, see Table S2), regardless

of whether we assume a single-pulse introgression model (Figure 2A) or two-pulse

introgression models (Figure S3), 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 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]

(Methods). Using this approach, we calculated the expected allele frequency in

four present day European samples from the 1,000 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 and colleagues. To set up our null

expectations we performed a similar analysis on a dataset of approximately one

million SNPs scattered around the genome, generated by Mathieson and

colleagues (by merging 213 ancient samples dated between 6,500 and 300 BCE

with sequencing data from four European samples from the 1,000 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 out of the 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, Figure 2B, Table S3). Our findings at this single locus are consistent with

results from the genome-wide selection scan performed by Mathieson and

colleagues [26] where the OAS region also showed evidence of selection ($P < 10^{-7}$

, 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 (Figure 3A and S4), further supporting that they have been targeted

by positive selection.

As an additional means of understanding whether or not 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 significantly longer than observed in simulated loci, and

that these results were robust to a series of alternative simulation models tested,

including the one- and two-pulse models described above and two additional one-

pulse models which varied the mutation and recombination rates (Figure 3B, Table

S4).

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 (Fst as high as 0.6, $P_{empirical} < 0.01$,

Figure 3C. 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

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 data collected on primary macrophages from 96

European-descent individuals, before and after *in-vitro* infection with *Salmonella*

typhimurium. After 2 hours of infection, we found that all OAS genes were

strongly up-regulated (up to 19-fold, $P < 1 \times 10^{-10}$, Figure S5), 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 *cis* eQTL for *OAS3*) (Figure

4A, false discovery rate (FDR) < 10%, Table S6). Interestingly, these *cis* eQTL

showed a much stronger effect in infected macrophages (best $P_{salmonella} = 3.5 \times 10^{-3}$

vs best $P_{non\text{-}infected} = 0.027$), 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-sequencing 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%; Figure 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 (A<u>G</u>

at acceptor site) retains the splice site whereas the derived allele, A, (A<u>A</u> at

acceptor site) disrupts the splice site leading to the usage of a distinct isoform

(Figure S6). 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 Neanderthal haplotype (with rare exceptions; ~2% of all haplotypes, Figure S2

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-sequencing data collected from

41 African-American individuals. As among Europeans, rs10774671 is a strong

asQTL for *OAS1* in both non-infected and infected macrophages ($P < 1 \times 10^{-15}$,

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 peripheral blood mononuclear cells

(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 in the OAS region (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 PCR (up to

27-fold, $P \leq 1.2 \times 10^{-8}$, Figure 5A), concomitantly with the up-regulation of type-I

and type-II interferon genes (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 ($P \leq 4.9 \times 10^{-5}$, Figure 5B). 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 =

6.1×10^{-3}) and the synthetic ligand gardiquimod (P = 2.0×10^{-4}), which mimics a

single strand RNA infection (Figure 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 signatures of positive selection in the European

population. Additionally, 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 Denosivan 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

(Figure S2). 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: *(i)* reduced infection and

replication rates of West Nile virus ([38], but see [39]), *(ii)* improved resistance to

hepatitis C virus (HCV) infection [40,41], and *(iii)* 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 Neandertal OAS haplotypes, 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 (Figure

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, 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 [43]. Further supporting

the idea of balancing selection on the introgressed haplotypes, 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 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 2,000-2,400 generations

ago (Methods). However, the observed allele frequency shift of 0.26 over the past

200-340 generations (maximum shift in CEU from ancient samples; see Figure

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 and

colleagues [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.

Materials and Methods

1. 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) [44]. 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/). For the frequency analysis, we

combined these filtered datasets with ten samples from outside of Africa (5-Eur

European: CEU, FIN, GBR, IBS, TSI; 5-East Asian: CDX, CHB, CHS, JPT,

KHB; see 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%2

06%20October%202014.html).

We first extracted all biallelic variants in our human, Neandertal, and chimpanzee

alignments. We considered as Neandertal-like sites (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 ($H_{D/A}$).

2. 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 5,008 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 one thousand 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.

3. 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 [45], 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 [46] in

order to explicitly simulate the genetic map (downloaded with the 1000 Genomes

samples at the link above) of the 600 kilobase region centered on OAS

(chr12:113100000-113700000). The general shape of the demographic model is

illustrated in Figure S1. Our baseline simulations were performed with the

parameters specified in Table S1, assuming 25 years per generation and a mutation

rate of 2.5×10^{-8} per bp per generation. Table S1 also provides all parameters used

in one- and two- introgression pulse simulations. Additionally, we examined the

one-pulse model with a slower mutation rate (1.25×10^{-8}) and a one-pulse model

with a uniform recombination rate (1.3×10^{-8}; a value conservatively lower than

the average for the OAS region, 1.7×10^{-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*. 2014 [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 >=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, 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 0.205198358413 -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

0.0641347992705 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_recrates.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 1 482.67144247 -eg 1e-12 2 460.409556336 -eg 2e-12 3 720.224280773 -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 1.98002735978 -eg

0.00699727402189 2 0.0 -eg 0.00699728402189 3 18.896348561 -en

0.03146374829 2 2.79399962717 -en 0.03146375829 3 0.670250141711 -ej

0.051785044418 3 2 -em 0.051785054418 1 2 0.00015 -em 0.051785064418 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

4. 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 (Figure 2A, S3). Minimum p-values across the OAS locus for

each population are listed in 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 [47],

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* [28,30,48]. 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

492 frequency, where they have a greater opportunity for recombination with non-

493 introgressed haplotypes, than neutral Neandertal haplotypes. We therefore defined

494 our test statistic, $H_{D/A}$, as:

495    $H_{D/A} = \ln(H_D / H_A)$;

496 where H_D and H_A are H calculated across haplotypes carrying derived NLS allele

497 versus the ancestral allele (see Figure S10).

498 For each model we generated 1,000 simulated OAS regions containing NLS that

499 reach a frequency greater than the 99th percentile for CEU in our frequency

500 simulations. We calculated $H_{D/A}$ for every simulated NLS and for every NLS in our

501 true sample. We then compared our simulation results to the observed data in two

502 ways. First, to visually examine the distribution of true $H_{D/A}$ values to neutral

503 simulations, we took the mean $H_{D/A}$ score across all NLS in each of 1,000

504 simulations and calculated the 95th and 99th percentiles of those 1,000 simulations

505 (Figure 3B). Second, to examine the probability under neutrality of the observed

506 $H_{D/A}$ values at the peak of $H_{D/A}$ between chr12:113380000-113420000, we

507 compared the mean $H_{D/A}$ score for SNPs in this window to the mean of SNPs in

508 this window for each simulation that produced a NLS in that window. These

509 results for all tested models are listed in Table S4.

510 *5. Analysis of ancient Eurasian data*

511 We utilized supplementary data table 3 from Mathieson *et al.* [26]. This table

512 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 1,000 Genomes Project (see

"Genome-wide scan for selection" section of methods in[26]). We intersect this

table with allele frequencies for 1,000 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 percent 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 Table

S3 and illustrated in Figure 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 Ps from that test to

ensure consistency with our results.

6. 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 (Δt), we assumed a model of standard logistic growth of a

codominant 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 Δt.

7. 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. [49]. This method estimates the probability that a haplotype of length (H) is

shared due to incomplete lineage sorting as:

P(H) = 1 - gammaCDF(H, shape = 2, rate = 1/L) ;

where L is the expected shared length, L = 1/(r x 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 x 10^{-8})

and a conservatively short divergence between Neandertals and humans of 300

thousand years (and 25 years per generation), the probability of a length of at least

54,385 is P(54385) = 0.002. The probability of the longest tracts is much smaller,

$P(198441) = 1.15 \times 10^{-12}$. 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 Figure 1A.

8. 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) and were filtered for

biallelic SNPs. Arlequin version 3.5.1.3 [50] 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 25kb either side of each SNP (i.e.,

in window of 50kb in total). iHS values were recovered from a genomic scan

performed using hapbin [51] on the same phase 3 release of the 1000 Genomes

dataset.

9. 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 project was approved by the ethics committee at the

CHU Sainte-Justine (protocol #4022). The individuals recruited in this study were

males aged 18 to 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 less than 55 years old), 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-

identifed ancestry labels, we used the genome-wide genotype data (see section 10)

to estimate genome-wide levels of European and African ancestry in each sample

using the program ADMIXTURE [52]. Consistent with previous reports, we found

that self-identified EA 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.

10. 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 [53]. Here, we only focused on

genetic diversity surrounding the OAS region – chr12:113229549-113574044 (~

344Kb) 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 Table S8.

11. Isolation of monocytes and differentiation of macrophages

Blood mononuclear cells were isolated by Ficoll-Paque centrifugation. Monocytes

were purified from peripheral blood mononuclear cells (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 7 days in RPMI-1640 (Fisher) supplemented with 10%

heat-inactivated FBS (FBS premium, US origin, Wisent), L-glutamine (Fisher)

and M-CSF (20ng/mL; R&D systems). Cell cultures were fed every 2 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)).

12. 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 *Salmonella 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 hours 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 hours, the time point to which we refer in

the main text.

13. 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 Neandertal origin and that we used as a proxy to identify individuals harbouring

the Neandertal haplotype in the OAS region. From the 40 individuals, and based

on this SNP, 9 individuals were homozygous for the Neandertal 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 (0.5μg/ml, TLR7 and TLR8

agonist), Influenza PR8 WT (multiplicity of infection (MOI) of 0.05:1), Herpes

simplex virus (HSV) 1 (1.55×10^2 CPE), and HSV2 (19.5×10^4 CPE). PBMCs were

stimulated/infected for 4 hours with TLR ligands and Influenza, and 6h with

HSV1 and HSV2. A control group of non-infected PBMC was treated the same

way but with only medium.

14. 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-

sequencing libraries were prepared using the Illumina TruSeq protocol. Once

prepared, indexed cDNA libraries were pooled (6 libraries per pool) in equimolar

amounts and sequenced with single-end 100bp reads on an Illumina HiSeq2500.

Results based on the entire dataset are described elsewhere (Nédélec *et al*., under

revision). 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 Table S9.

15. 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.

16. Quantitative real time PCR

For the PBMC samples we measured the expression levels of OAS and interferon

genes using real time PCR. 100ng of high-quality RNA was reverse-transcribed

into cDNA using the qScript cDNA SuperMix (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 2X 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). 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 2X 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 was analysed 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.

17. Genotype–Phenotype Association Analysis

eQTL, 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 false discovery rates (FDR) using

an approach analogous to that of Storey and Tibshirani[54], 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 Figure 4).

Author 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.

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Availability of data and materials: The datasets supporting the conclusions of this article are included within the article and in supplementary Tables 8 and 9. The raw data were deposited under GEO accession numbers GSE73765 and GSE81046.

Competing interests: The authors declare that they have no competing interests.

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Figures:

Figure 1: Neandertal introgresssed haplotypes in the OAS region. (A) Neighbor-

joining tree of 5008 phased haplotypes spanning chr12: 113344739-113449528

(hg19) from phase 3 of the 1000 Genomes Project. Haplotypes were condensed

into 10 core haplotypes based on the majority allele after clustering into groups of

haplotypes with pairwise differences of 60 or less. The figure illustrates that the

Altai haplotype is very similar to "cluster 2" haplotypes found in several human

populations, while the Denisovan haplotype is not more closely related to any of

the remaining (non-cluster 2) sequences in this dataset. Bootstrap values (1000

replicates) are provided in blue boxes at each node. (B) Frequencies of the 10 core

haplotypes within each 1000 Genomes Project population sample. The most

common Neandertal-like haplotype, cluster 2, is found only outside of sub-

Saharan African samples, with the exception of recently admixed populations.

Population codes can be found in Table S7.

Figure 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 one million SNPs scattered around the genome and with

comparable present-day frequencies to those found for NLS in the OAS region.

Figure 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 (grey) 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 1,000

simulations. (C) Fst calculated between CEU and all East Asian populations from

the 1000 Genomes projects (y-axis). Dashed lines indicate the 95th (grey) and 99th

(black) genome-wide percentiles of Fst. Similar results are obtained when

comparing any other European population against East Asians (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.

Figure 4: Pervasive impact of the Neandertal haplotype on the regulation of *OAS*

genes in primary macrophages. (A) –log 10 Ps (y-axis) for the association between

genotypes for SNPs with a MAF > 10% in the OAS region and expression levels

of *OAS3* in non-infected (black) and Salmonella-infected macrophages (orange).

The dashed line shows the P cutoff corresponding to an FDR of 10%. The right

panel shows a boxplot for the association between genotypes at the NLS

rs4767031 (x-axis) and the expression levels of *OAS3* in TPM (Transcripts Per

Kilobase Million) (y-axis). The lower expression level of *OAS3* in individuals

harboring the Neandertal haplotype were confirmed by real-time PCR (Figure

S11) (B) –log 10 Ps (y-axis) for the association between genotypes in the OAS

regions and the percentage usage of isoform ENST00000202917 (i.e., p46 in the

text) in non-infected (black) and Salmonella-infected macrophages (orange). The

dashed line shows the P cutoff corresponding to an FDR of 1%. The right panel

shows a boxplot for the association between genotypes at the splicing variant

rs10774671 (x-axis) and the percentage usage of isoform p46 (y-axis). (C) Similar

824   to (B) but for the percentage usage of isoform ENST00000449768 of OAS2. In all

825   the panels NLS are highlighted by blue dots. The arrows on panel A-C highlight

826   the location of the SNPs for which the boxplots are shown on the right.

Figure 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) –log10 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 Figure 3 as

associated with NLS). Boxplots show the association between genotype status

(blue referring to Neanderthal alleles) and expression levels of the p46 isoform of

OAS1 and *OAS3* in Gardiquimod-stimulated PBMCs.

852   **Supplementary Figures:**

854   **Figure S1:** Schematic representation of demographic model used in simulations.

855   Red arrows indicate points of introgression. The full range of demographic

856   parameters used for the simulation can be found in Table S1.

Figure S2: Haplogram of region surrounding *OAS1* splice site (rs10774671). 198

phased haplotypes from the CEU population sample are illustrated along with the

Altai and Denisovan haplotypes. Ancestral alleles are colored white, derived

alleles are (1) yellow if derived in Denisovan and absent from Africans (YRI)

(DLS), (2) red if derived in Neandertals and absent from Africans (YRI) (NLS),

(3) black otherwise. With the exception of two rare haplotypes in cluster 1, all

occurrences of the ancestral variant at rs10774671 are surrounded by derived

Neandertal alleles, indicating that these were introduced in a Neandertal

haplotype. Only sites where the Neandertal and Denisovan genomes are

homozygous were used in this haplogram. Cluster labels (y-axis) correspond to

clusters in Figure 1.

Figure S3: OAS-introgressed haplotypes are found at higher frequencies in

European populations than expected under neutrality when using a two-pulse

introgression model. (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 considering

two-pulse introgression in both Europeans and Asians, or a second pulse of

introgression only in Europeans. Dashed lines overlap almost completely,

reflecting little variation between a two-pulse introgression in both European and

Asians, or a second pulse only in Europeans.

887   **Figure S4:** The DIND statistic indicates reduced diversity surrounding NLS.

888   DIND (y-axis) is plotted for SNPs in OAS locus by derived (Neandertal) allele

889   frequency. Dashed line indicates genome-wide 95th percentile by frequency.

Figure S5: Expression levels of OAS genes in primary macrophages (European)

before and after infection with *Salmonella*.

Figure S6: The splicing variant rs10774671 is a strong asQTL for *OAS1*. Plotted

is the normalized average coverage at which each base was sequenced along the

genomic regions encoding the gene *OAS1*. Individuals were stratified according to

their genotype at rs10774671. Below the figure are gene models from the Ensembl

database. Individuals carrying the G allele at rs10774671 (i.e. the Neandertal

allele) primarily express the transcript ENST00000202917 (referred to as p46 in

the text) whereas individuals carrying the A derived allele lose the splice site,

which leads to the usage of a distinct isoform.

Figure S7: Isoform usage for transcript ENST00000202917 stratified by

genotype at rs10774671 in cells derived from individuals with African ancestry.

The derived (Neandertal) allele leads to increased expression of

ENST00000202917 in both non-infected and *Salmonella* infected conditions.

920   **Figure S8:** Log 2 fold induction (y-axis) of IFNA1 (type 1 interferon) and IFNG

921   (type II interferon) in PBMCs upon stimulation with several viral agents.

926   **Figure S9:** $H_{D/A}$ plotted against quintiles of segregating sites in 1,000 simulations.

927   The $H_{D/A}$ statistic does vary with number of segregating sites across simulated data.

Figure S10: Schematic illustration of the $H_{D/A}$ statistic. At each NLS, haplotypes

are divided into two groups, based on whether they carry the Neandertal (derived)

or African (ancestral) state. Within each haplotype subset, haplotype

homozygosity is calculated across all pairs of haplotypes and then averaged (H),

resulting in two values; H_A and H_D (A-ancestral, D-derived). Finally, $H_{D/A}$ is

calculated as the natural log of the ratio of derived to ancestral H values.

Excessively high values of this statistic reflect particularly long haplotypes

carrying Neandertal-derived alleles.

Figure S11: Boxplot for the association between genotypes at the NLS rs4767031

(x-axis) and the expression levels of *OAS3* in *Salmonella*-infected macrophages

using real-time PCR.

References:

1. Green RE, Krause J, Briggs AW, Maricic T, Stenzel U, Kircher M, et al. A draft
sequence of the Neandertal genome. Science. American Association for the
Advancement of Science; 2010;328:710–22.

2. Reich D, Green RE, Kircher M, Krause J, Patterson N, Durand EY, et al.
Genetic history of an archaic hominin group from Denisova Cave in Siberia.
Nature. Nature Publishing Group; 2010;468:1053–60.

3. Meyer M, Kircher M, Gansauge MT, Li H, Racimo F, Mallick S, et al. A high-
coverage genome sequence from an archaic denisovan individual. Science.
American Association for the Advancement of Science; 2012;338:222–6.

4. Prüfer K, Racimo F, Patterson N, Jay F, Sankararaman S, Sawyer S, et al. The
complete genome sequence of a Neanderthal from the Altai Mountains. Nature.
2014;505:43–9.

5. Sawyer S, Renaud G, Viola B, Hublin J-J, Gansauge M-T, Shunkov MV, et al.
Nuclear and mitochondrial DNA sequences from two Denisovan individuals.
Proceedings of the National Academy of Sciences. National Acad Sciences;
2015;:201519905.

6. Sankararaman S, Mallick S, Dannemann M, Prüfer K, Kelso J, Pääbo S, et al.
The genomic landscape of Neanderthal ancestry in present-day humans. Nature.
2014;507:354–7.

7. Llorente MG, Jones ER, Eriksson A, Siska V, Arthur KW, Arthur JW, et al.
Ancient Ethiopian genome reveals extensive Eurasian admixture throughout the
African continent. Science. 2015.

8. Vernot B, Akey JM. Resurrecting surviving Neandertal lineages from modern
human genomes. Science. 2014;343:1021.

9. Harris K, Nielsen R. The Genetic Cost of Neanderthal Introgression. Genetics.
Genetics; 2016;:genetics.116.186890.

10. Racimo F, Sankararaman S, Nielsen R, Huerta-Sanchez E. Evidence for
archaic adaptive introgression in humans. Nat Rev Genet. Nature Publishing
Group; 2015.

11. Racimo F, Marnetto D, Huerta-Sanchez E. The landscape of uniquely shared
archaic alleles in present-day human populations. bioRxiv. Cold Spring Harbor
Labs Journals; 2016;:045237.

12. Hublin JJ. The origin of Neandertals. Proceedings of the National Academy of

Sciences. 2009;106:16022–7.

13. Ségurel L, Quintana-Murci L. Preserving immune diversity through ancient
inheritance and admixture. Curr. Opin. Immunol. 2014;30C:79–84.

14. Mendez FL, Watkins JC, Hammer MF. A Haplotype at STAT2 Introgressed
from Neanderthals and Serves as a Candidate of Positive Selection in Papua New
Guinea. The American Journal of Human Genetics. Elsevier; 2012;91:265–74.

15. Mendez FL, Watkins JC, Hammer MF. Neandertal origin of genetic variation
at the cluster of OAS immunity genes. Molecular Biology and Evolution. Oxford
University Press; 2013;30:798–801.

16. Player MR, Torrence PF. The 2–5 A system: Modulation of viral and cellular
processes through acceleration of RNA degradation. Pharmacology &
Therapeutics. 1998;78:55–113.

17. Mendez FL, Watkins JC, Hammer MF. Global genetic variation at OAS1
provides evidence of archaic admixture in Melanesian populations. Molecular
Biology and Evolution. 2012;29:1513–20.

18. Deschamps M, Laval G, Fagny M, Itan Y, Abel L, Casanova J-L, et al.
Genomic Signatures of Selective Pressures and Introgression from Archaic
Hominins at Human Innate Immunity Genes. American Journal of Human
Genetics. 2016;98:5–21.

19. Sankararaman S, Mallick S, Patterson N, Reich D. The Combined Landscape
of Denisovan and Neanderthal Ancestry in Present-Day Humans. Current Biology.
2016.

20. Consortium T1GP. A global reference for human genetic variation. Nature.
Nature Publishing Group; 2015;526:68–74.

21. Fu Q, Li H, Moorjani P, Jay F, Slepchenko SM, Bondarev AA, et al. Genome
sequence of a 45,000-year-old modern human from western Siberia. Nature.
2014;514:445–9.

22. Yang MA, Malaspinas AS, Durand EY, Slatkin M. Ancient structure in Africa
unlikely to explain Neanderthal and non-African genetic similarity. Molecular
Biology and Evolution. SMBE; 2012;29:2987–95.

23. Gravel S, Henn BM, Gutenkunst RN, Indap AR, Marth GT, Clark AG, et al.
Demographic history and rare allele sharing among human populations.
Proceedings of the National Academy of Sciences of the United States of America.
National Acad Sciences; 2011;108:11983–8.

24. Tennessen JA, Bigham AW, O'Connor TD, Fu W, Kenny EE, Gravel S, et al.
Evolution and Functional Impact of Rare Coding Variation from Deep Sequencing

of Human Exomes. Science. 2012;337:64–9.

25. Vernot B, Akey JM. Complex History of Admixture between Modern Humans
and Neandertals. The American Journal of Human Genetics. 2015.

26. Mathieson I, Lazaridis I, Rohland N, Mallick S, Patterson N, Roodenberg SA,
et al. Genome-wide patterns of selection in 230 ancient Eurasians. Nature. 2015.

27. Lazaridis I, Patterson N, Mittnik A, Renaud G, Mallick S, Kirsanow K, et al.
Ancient human genomes suggest three ancestral populations for present-day
Europeans. Nature. Nature Publishing Group; 2014;513:409–13.

28. Voight BF, Kudaravalli S, Wen X, Pritchard JK. A map of recent positive
selection in the human genome. Plos Biology. Public Library of Science;
2006;4:e72.

29. Barreiro LB, Ben-Ali M, Quach H, Laval G, Patin E, Pickrell JK, et al.
Evolutionary dynamics of human Toll-like receptors and their different
contributions to host defense. Mcvean G, editor. PLoS Genetics. Public Library of
Science; 2009;5:e1000562.

30. Sabeti, P.C., Reich DE, Higgins JM, Levine HZP, Richter DJ, Schaffner SF, et
al. Detecting recent positive selection in the human genome from haplotype
structure. Nature. Nature Publishing Group; 2002;419:832–7.

31. Nauciel C, Espinasse-Maes F. Role of Gamma Interferon and Tumor-Necrosis-
Factor-Alpha in Resistance to Salmonella-Typhimurium Infection. Infect. Immun.
American Society for Microbiology; 1992;60:450–4.

32. Gulig PA, Doyle TJ, Clare-Salzler MJ, Maiese RL, Matsui H. Systemic
infection of mice by wild-type but not Spv- Salmonella typhimurium is enhanced
by neutralization of gamma interferon and tumor necrosis factor alpha. Infect.
Immun. American Society for Microbiology (ASM); 1997;65:5191–7.

33. LaRock DL, Chaudhary A, Miller SI. Salmonellae interactions with host
processes. Nat Rev Micro. 2015;13:191–205.

34. Wu L, Candille SI, Choi Y, Xie D, Jiang L. Variation and genetic control of
protein abundance in humans. Nature. 2013;499:79–82.

35. Pickrell JK, Marioni JC, Pai AA, Degner JF. Understanding mechanisms
underlying human gene expression variation with RNA sequencing. Nature.
2010;464:768–72.

36. Bonnevie-Nielsen V, Field LL, Lu S, Zheng D-J, Li M, Martensen PM, et al.
Variation in antiviral 2',5'-oligoadenylate synthetase (2"5"AS) enzyme activity is
controlled by a single-nucleotide polymorphism at a splice-acceptor site in the
OAS1 gene. The American Journal of Human Genetics. 2005;76:623–33.

37. Meyer M, Arsuaga JL, de Filippo C, Nagel S, Aximu-Petri A, Nickel B, et al.
Nuclear DNA sequences from the Middle Pleistocene Sima de los Huesos
hominins. Nature. 2016.

38. Lim JK, Lisco A, McDermott DH, Huynh L, Ward JM, Johnson B, et al.
Genetic Variation in OAS1 Is a Risk Factor for Initial Infection with West Nile
Virus in Man. Rice CM, editor. PLoS Pathog. Public Library of Science;
2009;5:e1000321.

39. Bigham AW, Buckingham KJ, Husain S, Emond MJ, Bofferding KM,
Gildersleeve H, et al. Host Genetic Risk Factors for West Nile Virus Infection and
Disease Progression. Ravichandran V, editor. PLoS ONE. Public Library of
Science; 2011;6:e24745.

40. Kwon Y-C, Kang J-I, Hwang SB, Ahn B-Y. The ribonuclease l-dependent
antiviral roles of human 2′,5′-oligoadenylate synthetase family members against
hepatitis C virus. FEBS Letters. 2012;587:156–64.

41. Awady El MK, Anany MA, Esmat G, Zayed N, Tabll AA, Helmy A, et al.
Single nucleotide polymorphism at exon 7 splice acceptor site of OAS1 gene
determines response of hepatitis C virus patients to interferon therapy. Journal of
Gastroenterology and Hepatology. Blackwell Publishing Asia; 2011;26:843–50.

42. Barreiro LB, Quintana-Murci LIS. From evolutionary genetics to human
immunology: how selection shapes host defence genes. Nat Rev Genet. Nature
Publishing Group; 2010;11:17–30.

43. Hancks DC, Hartley MK, Hagan C, Clark NL, Elde NC. Overlapping Patterns
of Rapid Evolution in the Nucleic Acid Sensors cGAS and OAS1 Suggest a
Common Mechanism of Pathogen Antagonism and Escape. Zhang J, editor. PLoS
Genetics. Public Library of Science; 2015;11:e1005203.

44. Kent WJ, Sugnet CW, Furey TS, Roskin KM, Pringle TH, Zahler AM, et al.
The human genome browser at UCSC. Genome Research. Cold Spring Harbor
Lab; 2002;12:996–1006.

45. Hudson R. Generating samples under a Wright–Fisher neutral model of genetic
variation. Bioinformatics. 2002.

46. Chen GK, Marjoram P, Wall JD. Fast and flexible simulation of DNA
sequence data. Genome Research. 2009;19:136–42.

47. Schlamp F, Made J, Stambler R, Chesebrough L, Boyko AR, Messer PW.
Evaluating the performance of selection scans to detect selective sweeps in
domestic dogs. Mol Ecol. 2016;25:342–56.

48. Ferrer-Admetlla A, Liang M, Korneliussen T, Nielsen R. On detecting
incomplete soft or hard selective sweeps using haplotype structure. Molecular

Biology and Evolution. Oxford University Press; 2014;31:1275–91.

49. Huerta-Sanchez E, Jin X, Asan, Bianba Z, Peter BM, Vinckenbosch N, et al.
Altitude adaptation in Tibetans caused by introgression of Denisovan-like DNA.
Nature. 2014;512:194–7.

50. Excoffier L, Lischer HEL. Arlequin suite ver 3.5: a new series of programs to
perform population genetics analyses under Linux and Windows. Molecular
Ecology Resources. Blackwell Publishing Ltd; 2010;10:564–7.

51. Maclean CA, Chue Hong NP, Prendergast JGD. hapbin: An Efficient Program
for Performing Haplotype-Based Scans for Positive Selection in Large Genomic
Datasets. Molecular Biology and Evolution. Oxford University Press;
2015;32:3027–9.

52. Alexander DH, Novembre J, Lange K. Fast model-based estimation of
ancestry in unrelated individuals. Genome Research. Cold Spring Harbor Lab;
2009;19:1655–64.

53. Howie BN, Donnelly P, Marchini J. A Flexible and Accurate Genotype
Imputation Method for the Next Generation of Genome-Wide Association Studies.
Schork NJ, editor. PLoS Genetics. Public Library of Science; 2009;5:e1000529.

54. Storey JD, Tibshirani R. Statistical significance for genomewide studies.
Proceedings of the National Academy of Sciences of the United States of America.
National Acad Sciences; 2003;100:9440–5.

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