<u>Dual RNA-seq provides insight into the biology of the neglected intracellular human pathogen *Orientia*</u>

<u>*tsutsugamushi*</u>

Bozena Mika-Gospodorz[#a], Suparat Giengkam[#b], Alexander J. Westermann[a,c], Jantana Wongsantichon[b],

Willow Kion-Crosby[d], Suthida Chuenklin[b], Loo Chien Wang[e], Piyanate Sunyakumthorn[f], Radoslaw M.

Sobota[j], Selvakumar Subbian[h], Jörg Vogel[a,c], Lars Barquist[#*a,g] and Jeanne Salje[#*b,h,i]

Running title: RNA sequencing of *Orientia tsutsugamushi*

Helmholtz Institute for RNA-based Infection Research (HIRI), Helmholtz Centre for Infection Research

(HZI), Würzburg, Germany[a]

Mahidol-Oxford Tropical Medicine Research Unit, Faculty of Tropical Medicine, Mahidol University,

Bangkok, Thailand[b];

Institute for Molecular Infection Biology (IMIB), University of Würzburg, Würzburg, Germany[c]

Rutgers, the State Univeristy of New Jersey, New Jersey, USA[d]

Functional Proteomics Laboratory, Institute of Molecular and Cell Biology, Agency for

Science, Technology and Research (A*STAR), Singapore[e].

Armed Forces Research Institute of Medical Sciences, Bangkok, Thailand[f];

Faculty of Medicine, University of Würzburg, Würzburg, Germany[g];

Public Health Research Institute, Rutgers University, New Jersey, USA[h];

Centre for Tropical Medicine and Global Health, Nuffield Department of Medicine, University of Oxford,

Oxford, United Kingdom[i];

SingMass - National Mass Spectrometry Laboratory, Institute of Molecular and Cell Biology, Agency for

Science, Technology and Research (A*STAR), Singapore[j]

[#] these authors contributed equally

*address correspondence to js2522@njms.rutgers.edu; lars.barquist@helmholtz-hiri.de

Summary

Emerging and neglected diseases pose challenges as their biology is frequently poorly understood, and genetic tools often do not exist to manipulate the responsible pathogen. Organism agnostic sequencing technologies offer a promising approach to understand the molecular processes underlying these diseases. Here we apply dual RNA-seq to *Orientia tsutsugamushi* (Ot), an obligate intracellular bacterium and the causative agent of the vector-borne human disease scrub typhus. Half the Ot genome is composed of repetitive DNA, and there is minimal collinearity in gene order between strains. Integrating RNA-seq, comparative genomics, proteomics, and machine learning, we investigated the transcriptional architecture of Ot, including operon structure and non-coding RNAs, and found evidence for wide-spread post-transcriptional antisense regulation. We compared the host response to two clinical isolates and identified distinct immune response networks that are up-regulated in response to each strain, leading to predictions of relative virulence which were confirmed in a mouse infection model. Thus, dual RNA-seq can provide insight into the biology and host-pathogen interactions of a poorly characterized and genetically intractable organism such as Ot.

Keywords: neglected and emerging pathogens, intracellular bacteria, dual RNA-seq, transcriptomics, host-pathogen cell biology, bacterial virulence, antisense transcription.

Main Text

Introduction

Improved surveillance and diagnostics have led to the recognition of previously neglected bacteria as

serious pathogens, whilst human population growth, globalization and increased travel have contributed

to the emergence of new pathogens and changing patterns of infectious disease. The biology of neglected

and emerging pathogens is often poorly understood, but is essential to developing therapeutic and

preventative strategies. Obligate intracellular pathogens present additional challenges, as many cause

diseases that are difficult to diagnose and are difficult to manipulate experimentally.

Obligate intracellular bacteria include the Rickettsiales, an order which includes the arthropod and

nematode symbiont *Wolbachia* as well as a number of human and veterinary pathogens. *Orientia*

tsutsugamushi (Ot, Class Alphaproteobacteria, Order Rickettsiales, Family Rickettsiaceae) causes the

mite-borne human disease scrub typhus, a leading cause of severe febrile illness in the Asia Pacific region[1],

home to roughly two thirds of the world's population. Locally acquired cases in the Middle East and Latin

America suggest that this disease may be more widespread than previously appreciated[2,3]. Under-

recognition and under-reporting are a major problem in scrub typhus because unambiguous diagnosis is

difficult, and awareness is low amongst many clinicians. Symptoms are non-specific and include

headache, fever, rash, and lymphadenopathy beginning 7-14 days after inoculation via a feeding larval

stage mite. If untreated, this can progress to cause multiple organ failure and death. In the mite vector Ot

colonizes the ovaries and salivary glands. During acute infection of its mammalian host, the bacteria infect

endothelial cells, dendritic cells and monocytes/macrophages at the mite bite site[4], and then disseminate

via blood and lymphatic vessels to multiple organs including lung, liver, kidney, spleen and brain[5].

Ot strains are highly variable in terms of antigenicity and virulence. Hundreds of strains have been

described based on differences in the sequence of the surface protein TSA56[9–11]. These strains are

classified into 7 geographically diverse serotype groups, dominated by the Karp, Kato and Gilliam

groups[12]. As with many other pathogenic bacteria, whole genome sequencing has revealed that serotype-

based groupings do not necessarily reflect phylogenetic relationships[8]. Different strains of Ot exhibit

different levels of virulence[19–21], dependent on both bacterial and host genotype. For example, strain Karp

(serotype group Karp) causes lethal infection in BALBc and C3H/He mice at low doses, strain Gilliam

(serotype group Gilliam) causes lethal infection in C3H/He but not BALBc mice at similar doses, whilst

strain TA716 (serotype group TA716) does not cause lethal infection in either mouse model at similar

doses[20,22]. The underlying causes of this variation in infection outcomes remain obscure.

Dual RNA-seq quantifies RNA transcripts of intracellular pathogens and host cells in a single

experiment[24,25], and can provide insight into both the host and pathogen response to infection. For

example, dual RNA-seq has been used to study obligate intracellular *Chlamydia trachomatis*[26] revealing

the rewiring of *Chlamydia* metabolism during the onset of an infection of human epithelial cells, together

with the corresponding host responses. Here we apply dual RNA-seq to deepen our understanding of the

RNA biology of Ot and its consequences for virulence. We survey the transcriptome of Ot strain Karp,

identifying non-coding RNAs and transcribed operons in a genome broken by frequent recombination and

transposition of the <u>r</u>ickettsial <u>a</u>mplified <u>g</u>enetic <u>e</u>lement (RAGE) integrative and conjugative element

(ICE)[6,7]. Integrating proteomic measurements, we further provide evidence that RAGE genes are regulated

through prevalent antisense transcription. Finally, we compare infection between strain Karp and strain

UT176 identifying a core host response to Ot dominated by type-I interferon signaling, as well as distinct

immune responses to each strain. We show that this in turn leads to different outcomes in a mouse model

of scrub typhus.

Results

Dual RNA-seq of *Orientia tsutsugamushi* infecting endothelial cells

We focused on two Ot clinical isolates: Karp, taken from a patient in New Guinea in 1943[27], and UT176, a separate strain in the same serotype group as Karp, taken from a patient in northern Thailand in 2004[28]. These strains are closely related, with a sequence identity of 95% in their TSA56 gene (used to classify strains)[8]. Consistent with a closed pan-genome for Ot, the gene content of Karp and UT176 are similar, with differences primarily in gene copy number, pseudogenes, and gene order along the genome. Human vein endothelial cells (HUVEC) were selected as host cells due to their similarity to cell types involved in both early and advanced infection. HUVEC cells were infected with bacteria at an MOI of 100:1 and grown for 5 days (Fig. 1A), by which point host cells were heavily loaded with bacteria (Fig. 1C, Supp. Figs 1, 2). Uninfected HUVEC cells were grown in parallel. After 5 days total RNA was isolated, depleted for rRNA, converted to cDNA and sequenced to ~35 million reads per library using Illumina technology. Reads were mapped to the completed genomes of Karp, UT176[8] and, in parallel, the human genome. As the *Orientia* genome is repeat-rich, we additionally applied model-based quantification with Salmon[29] which uses uniquely mapping reads to assign multi-mapping reads to these transcriptomes to improve our estimates of transcript abundance (see Methods).

We observed 17.1-17.5% bacterial reads in HUVECs infected with Karp and 2.8-4.9% bacterial reads in HUVECs infected with UT176 (Fig. 1D, Supp. Fig. 3). This difference likely reflects growth rate differences between Karp and UT176, which have doubling times of 19 and 27 hours in HUVEC, respectively (Fig. 1B). The distribution of reads to RNA classes (Fig. 1D) indicated efficient depletion of ribosomal transcripts in the host transcriptome (<0.001% human rRNA reads). In contrast, we found an average of 16% and 30% rRNA reads in Karp and UT176, respectively. Most of these remaining bacterial

ribosomal reads were derived from 5S rRNA (Supp. Table), likely reflecting the divergence of 5S rRNA sequences between Ot and bacterial model organisms used for optimization of the Ribo-Zero approach (https://emea.illumina.com/products/selection-tools/ribo-zero-kit-species-compatibility.html?langsel=/de/). This notwithstanding, reads mapping to coding sequences (CDSs) were not only abundant in the HUVEC data subset (53% of all host-mapped reads), but also in the Ot-specific reads (38% of the Karp- and 49% of the UT176-mapped reads), allowing differential expression analysis. Dual RNA-seq also readily detected the various non-coding RNA classes from both host and bacteria (Fig. 1E). Of 657 predicted core Ot genes[8] 491 were expressed and 212 were highly expressed (see Methods for definitions).

Ot ncRNAs and evidence for tmRNA processing

Bacterial genomes encode many non-coding (nc)RNAs. Among the most conserved are several specialized, abundant housekeeping ncRNAs, including the RNA components of ribonuclease P (RNase P), the signal recognition particle (SRP), and transfer-messenger RNA (tmRNA), all of which were detected in the Karp transcriptome data (Fig. 1E, Supp. Table 1). To validate the RNA-seq data, we performed Northern blot analysis for conserved housekeeping ncRNAs (Fig. 2A). These include the M1 RNA component of RNase P, a ribozyme responsible for tRNA processing, and 4.5S, the RNA component of the SRP involved in translocation of membrane proteins. Both ran at their expected lengths of ~385 and ~100 nt, respectively. However, a second stronger band for the M1 transcript ran slightly higher, indicative of a length of ~450 nt, suggesting the existence of a precursor-M1.

We also found evidence of tmRNA processing in Ot. tmRNA has both mRNA-like and tRNA-like features, rescues stalled ribosomes[30], and is known to contribute to virulence in pathogens as diverse as

Salmonella Typhimurium[31] and *Francisella tularensis*[32]. In our data, tmRNA appears to be expressed at

unusually high levels, contributing between 3 and 7% of total bacterial reads (Fig 1E), suggesting an

important role in Ot survival in mammalian cells. tmRNA generally consists of a tRNA-like (acceptor)

domain encoded upstream of a short open reading frame (coding domain). However, the transcript has

undergone a circular permutation in some clades of bacteria[33], including the Alphaproteobacteria[34], which

requires processing of a precursor transcript into separate, base-pairing acceptor and coding RNA

chains[35,36] (Fig. 2B). We detected three Ot tmRNA forms using Northern blot: (i) a long precursor tmRNA

(372 nt); (ii) a 5' fragment of ~80 nt, the acceptor domain; (iii) and the 3' coding domain of ~240 nt (Fig.

2B). Read coverage over the tmRNA locus in the Karp genome supported a cleavage event within the

loop region that connects the tRNA- and mRNA-like domains in the full-length precursor (Fig. 2C).

In addition to these universally conserved housekeeping ncRNAs, bacterial genomes encode family-,

genus-, species-, or strain-specific small ncRNAs (sRNAs) to adapt their gene expression to specific

intrinsic and environmental cues[37,38]. Our RNA-seq data identified 55 intergenic sRNA candidates,

between 77-803 nt, in the Karp transcriptome (Supp. Table 1). When normalized to the genome size of

Ot, this is consistent with the number of sRNAs reported in model bacterial pathogens[39–44].

Conserved operons in a dynamic genome

The genome of Ot is highly dynamic[8], and while the timescales and mechanisms of its rearrangements are

unknown they are thought to be driven by an extreme proliferation of mobile elements[6,7], in particular the

RAGE. The consequences of this are evident when comparing the high degree of synteny in bacteria from

two related 'normal' genera (*Escherichia* and *Salmonella*) to the complete shuffling we observe between

the two Ot strains studied here (Fig. 2D). As bacterial genomes are normally organized into co-transcribed

operons of functionally related genes, we wondered how this macroscale loss of synteny would affect

conservation of these transcripts. Using Rockhopper[45] and manual curation, we identified adjacent genes

expressed in a continuous transcript, classifying these as operons. We identified 131 operons fully

conserved between Karp and UT176 (all genes expressed in both strains) and seven partly conserved

(some genes expressed in both strains). Our previous analysis of 8 Ot genomes identified 51 universally

conserved genomic islands, including 35 potential collinear gene clusters containing two to thirteen

genes[8], and we found evidence for operonic transcripts originating from 24 of these. We also identified

212 and 192 transcribed operons present only in Karp or UT176, respectively, and these were generally

associated with the RAGE mobile element (73% in Karp and 93% in UT176) in contrast to conserved

operons (14% of fully conserved operons, Fig 2E).

The majority (84%, Supp. Fig. 6) of conserved operons consisted of only two or three genes. Longer

operons tended to encode for core cellular processes, the longest being a 30 gene operon encoding almost

half of Ot ribosomal proteins proximal to the ribosomal RNA operon itself (Fig. 2F). Others included an

8 gene operon involved in iron-sulfur cluster assembly, and 6 and 5 gene operons in distinct loci each

encoding for portions of the NADH-ubiquinone oxidoreductase complex in an organization similar to that

observed in *Rickettsia prowazekii* and eukaryotic mitochondria[46]. In summary, the identification of

cotranscribed gene clusters in a genome as highly dynamic as that of Ot indicates strong selection for

those genes to remain coupled, indicating involvement in the same pathways and likely shared regulation.

Evidence for Ot RAGE regulation by antisense RNA

The RAGE of Ot is present in at least 185 remnant copies[7]. It encodes an integrase (*int*) and transposase

gene *(tra),* multiple genes from the VirB type IV secretion system (*vir*) and a number of potential effector

genes including ankyrin-repeat containing proteins (*ank*), tetratricopeptide repeat containing proteins (*TPR*), SpoT/RelA genes, DNA methyltransferases and replicative DNA helicases. Many of these genes are truncated and most RAGE copies are highly degraded, containing only a subset of genes from the complete element. It is not known if this ICE is still active for transposition, nor whether Ot can express a functional type IV secretion apparatus. In our RNA-seq dataset ~50% of the most highly expressed genes were repetitive genes encoded by the RAGE (defined throughout our analysis as integrase, transposase, conjugal transfer genes and hypothetical genes) in both strains. These same genes were also highly expressed in the antisense direction (Fig. 3A, Supp. Fig. 6), leading us to hypothesize that the repetitive RAGE genes may be regulated by antisense gene expression.

Antisense transcription is widespread in bacteria[47], with between 5% and 75% of coding sequences exhibiting antisense transcription. While functions for a number of specific antisense transcripts have been described, including regulation through occlusion of the ribosome binding-site or induction of RNase III mediated decay[48] their relevance as a general functional class remains unclear. Antisense promoters tend to be weakly conserved[49], arguing against specific functions, and mathematical modeling has suggested the majority of antisense transcripts are not expressed at sufficient levels to affect the regulation of their cognate coding sequence[50].

To explore the relationship between sense and antisense expression of core Ot genes and repetitive RAGE genes, we combined our Karp RNA-seq dataset with a proteomics dataset generated under the same experimental growth conditions. We observed substantially fewer RAGE gene products detected by proteomics, compared with RNA-seq (Fig. 3A). Genes with detected protein products had higher transcript expression on average compared to those not detected by proteomics (Fig. 3B). However, many highly expressed transcripts appeared to produce no protein. Given our previous observations, we asked whether antisense transcription would correlate with protein expression. All genes with detected proteins

had an antisense-sense read count ratio of less than 1, in contrast to genes with no detected protein product, which had an antisense-sense read count spanning several orders of magnitude (Fig. 3C) suggesting antisense RNA expression may be a factor in inhibiting translation.

To test this hypothesis more rigorously, we constructed three logistic regression models to predict protein detection from our transcriptomic data. The first used only transcripts per million (TPMs) derived from the sense strand as a predictor; the second used only the antisense-sense read count ratio as a predictor; the third used both features. Comparisons of the predictive power of these three models showed that antisense transcription is predictive of protein expression (Fig. 3D). Model 1, relying only on sense expression, did little better than chance at predicting protein detection. Models 2 and 3, which incorporate the antisense-sense ratio, led to large improvements in predictive power, suggesting that antisense transcription plays a wide-spread regulatory role in Ot. This was confirmed by cross-validation (Methods, Supp. Fig. 7). We found significant enrichment for RAGE genes among those with high antisense-sense ratios (Fig. 3E), suggesting antisense transcription may work to control the expression of selfish genetic elements at the protein level. Thirty core genes also exhibited an antisense-sense ratio of >1 (Supp. Table core genes) and these include the chromosomal replication initiator protein DnaA, DNA polymerase subunit III, an outer membrane autotransporter protein ScaD, glutamine synthetase, three transporters, the protein export protein SecB and 12 hypothetical proteins. None of these models achieved greater than 67% balanced accuracy, which may be due to both the existence of other modes of post-transcriptional regulation and the lack of sensitivity in our proteomics. For instance, we have also performed a preliminary investigation of codon bias and found some evidence for differential codon usage in genes expressed at the RNA, but not protein, level (Supp. Text, Supp. Fig. 8-9).

Differential expression of genes in Karp and UT176

Due to a lack of genetic tools, identification of virulence mechanisms in Ot has been difficult, with only

a small number of antigenic surface proteins and effectors known. As pangenome diversity appears to

primarily be the result of gene duplication and decay, differences in virulence between strains are likely

due to differences in expression. To investigate this hypothesis, we performed differential expression

analysis between Karp and UT176 at 5 d after infection of HUVEC cells. Pathway and gene ontology

(GO) analyses of differentially expressed genes (Fig. 4A, Supp. Table 1) indicated that most pathways

were up-regulated in Karp compared with UT176, including those involved in DNA replication and

metabolism, consistent with Karp's higher growth rate (Fig. 1B). At the gene level (Supp. Table 1, Fig.

4B) we found a number of surface and effector proteins (Anks) were differentially regulated between the

two strains. Ot encodes five autotransporter domain-containing proteins (ScaA-ScaE) and three

immunogenic-type surface antigens (TSA22, TSA47, TSA56). All these surface proteins are

immunogenic, based on their reactivity to patient sera[60], with TSA56 being the most abundant Ot surface

protein. TSA56 has four variable domains and these lead to strain-specific antibody responses in patients.

TSA47, TSA56, and ScaA have been evaluated as possible vaccine candidates[61,62]. Of the core Ot genes,

those most differentially expressed between Karp and UT176 were *scaE*, *tsa56,* and *tsa22* (1.39, 3.06 and

3.94 logFC in Karp over UT176, respectively). In contrast, *scaD* levels were increased in UT176 but to a

lesser degree (0.80 logFC in UT176 over Karp). Given their high immunogenicity it is likely that their

differential expression results in differential host responses.

Genes for Ank and tetratricopeptide repeat-containing proteins (TPR) are present in 33 (Ank)/29 (TPR)

and 21 (Ank)/22 (TPR) copies in Karp and UT176, respectively [8]. Some *ank* genes function as effectors

in eukaryotic cells whilst others are uncharacterized. We compared the expression of Ank and TPR genes

in Karp and UT176. *ank2*, *ank3*, *ank12*, *tpr6,* and *tpr8* were up-regulated with a logFC >1.5 in UT176,

while *ank16, ank17, ak20,* and *tpr1* were up-regulated with a logFC >1.5 in Karp. Most of these proteins

were not detected in the Karp proteomics dataset suggesting that either the mRNAs were not translated,

or that the proteins were secreted and lost during purification. The protein products of all of these *ank*

genes localise to the endoplasmic reticulum or host cell cytoplasm when ectopically expressed[63]. Ank6

interferes with NFkB translocation to the nucleus and inhibits its transcriptional activation[64]. The activity

of the other differentially expressed Anks is not known. Given that these effector proteins interact directly

with host cell proteins we expect that this differential expression will lead to downstream differences in

host response.

Karp and UT176 induce a type-I interferon proinflammatory response

The transcriptional profile of HUVEC cells infected with Karp or UT176 showed a clear core response to

Ot (Fig. 5A, red), with smaller gene sets responding specifically to a single strain (Fig. 5A purple and

orange). The core response was dominated by a type-I interferon proinflammatory response (Supp. Table

1), seen previously in cultured endothelial cells and monocytes, as well as patient-derived macrophages[13–

17]. This is further illustrated by activation of the canonical interferon signaling pathway in response to

Karp (Fig. 5B), with a similar response observed for UT176 (Fig. 5C).

Host genes commonly up-regulated upon infection with either Ot strain include *IFNB1* (interferon beta)

and genes involved in regulating the type-I interferon response: *IRF9* (interferon-regulated factor 9) and

STAT1/2. Interferon-stimulated genes were also up-regulated upon Ot infection, including various

interferon induced proteins with tetratricopeptide repeats (*IFIT*) genes and 2'-5'-oligoadenylate synthase

1 (*OAS1*). In addition to the type-I interferon pathway, the joint Ot response led to up-regulation of

proinflammatory chemokine genes including *CXCL10*, *CXCL11*, and for cytokine receptors *IL13RA2*,

IL7R, *IL15RA*, and *IL3RA* (Supp. Table 1).

The upstream signals leading to activation of these signaling pathways are unknown but Ot has been

shown to activate host cells by signaling through the NOD1-IL32[65] and TLR2[66] pathways. Our data

showed that *TLR3* is up-regulated in cultured HUVEC cells in response to both Karp and UT176 (Supp.

Table 1). TLR3 recognizes viral double-stranded (ds)RNA in the cytoplasm[67] and it is possible that it

responds to Ot dsRNA. The up-regulation of the mRNA for transcription factor IRF7, which is known to

respond to stimulation from membrane-bound TLRs, further supports a role for TLR2 and TLR3 in the

detection of Ot.

Differential host responses to Karp and UT176

Although Karp and UT176 both induced a type-I interferon proinflammatory response compared to

uninfected HUVEC cells (Fig. 5B,C), each strain also induced its own unique response. Some of these

expression changes were validated by qRT-PCR (Supp. Fig. 12). The mRNA levels of multiple cytokines,

chemokines, and cytokine receptors were higher in HUVEC cells infected with UT176 compared with

Karp (Fig. 6A, Supp Fig 14, 15, 17). A network map of proinflammatory chemokines and cytokines, and

their differential induction in response to UT176 and Karp is shown in Fig 6A and Supp. Fig. 14A. Most

of the genes for cytokines, chemokines, and cytokine receptors were differentially up-regulated by

infection with UT176 compared with Karp, including *CXCL8*, *CXCL1*, *CXCL2*, *CXCL10*, *IL6*, *IL1RL1*,

IL18R1. The mRNA levels of surface adhesion molecules associated with activation of the endothelium,

VCAM1 and ICAM1, were also up-regulated in UT176 infected cells compared with Karp (Supp. Table

1). Whilst *TLR3* was up-regulated in both strains, TLR3 activation in UT176 infected cells was 1.5 logFC

higher than in response to Karp. Comparison of NFkB pathway genes and genes associated with NOS2 production revealed that genes in both pathways were up-regulated in UT176 but they were not up-regulated or were significantly less up-regulated in response to Karp infection (Supp. Fig. 15). Expression of host genes associated with leukocyte proliferation and mononuclear leukocyte differentiation was strongly induced in HUVECs infected with UT176 but significantly less so when infected with Karp (Supp. Fig. 16). Thus, UT176 seems to induce a stronger proinflammatory response and this may lead to more effective pathogen clearance (see Fig. 1B).

In contrast to the multiple chemokines and cytokines up-regulated in UT176-infected HUVEC cells, only *IL33* was specifically up-regulated in Karp-infected HUVEC cells (5 logFC difference; Supp. Table 1). IL33 is a proinflammatory cytokine that is involved in pathogenicity in a mouse model of scrub typhus[68]. To investigate Karp-mediated activation of *IL33* we analyzed gene induction in the IL33-FAS network (Fig. 6B/Supp. Fig. 14B). Most genes in the network were differentially induced in Karp-infected HUVEC cells compared to in UT176-infected HUVECs. Up-regulation of IL33-NOS-mediated signaling contributes to tissue inflammation. We analyzed networks of genes involved in (i) organismal growth failure (ii) organismal morbidity and mortality and (iii) organismal death. In all cases Karp induced these networks whilst UT176 dampened them (Supp. Fig. 17).

Two Ot strains differ in virulence in a mouse model

To assay the effects of the differential inflammatory responses induced by Karp and UT176 we tested the relative virulence of the two strains in an intravenous mouse infection model. 1.25×10^6 bacteria were inoculated into outbred ICR mice and monitored for disease symptoms for 12 days prior to euthanasia. Blood and tissue from lung, liver, spleen, and kidneys were isolated and the bacterial load measured by

322   qPCR. We found that Karp was significantly more virulent than UT176 as determined by disease outcome,

323   weight loss, bacterial load, and histopathological analysis (Fig. 7, Supp. Fig. 19-21). This difference is

324   likely to result from a combination of differential bacterial growth rate (Fig. 1B), differential expression

325   of bacterial virulence related genes (Fig. 4), and differential induction of host immune networks (Fig. 6).

326
327

Discussion

Both its obligate intracellular lifestyle and the complexity of rearrangements in the Ot genome make it difficult to study. Ot has a genome of 1.9-2.5 Mbp, almost half of which is composed of repetitive regions of >1,000 bp in length[8]. This is in contrast to the most closely related rickettsial species, whose genomes are typically around 1.1-1.3 Mbp[69]. The Ot genome is remarkably unstable, which makes inference of its transcriptional architecture particularly difficult. Using RNA-seq, we were able to identify core ncRNAs, putative sRNAs, and operonic transcripts. In sharp contrast to most bacteria, only a handful of operons containing more than 2 or 3 genes were conserved between Karp and UT176, and these primarily encode for proteins involved in core cellular processes like respiration and translation. Given that Karp encodes only 12 predicted transcription factors and 3 sigma factors, in contrast to 300 and 7, respectively, in *E. coli*, this raises the question of how transcription in Ot is coordinated.

One possible explanation is that much Ot transcription is not stringently controlled, and alternative mechanisms have arisen in Ot to control protein expression. This is supported in part by our observation that protein expression is partially predicted by antisense transcription. This mode of regulation seems to be particularly prevalent for genes encoded by the RAGE, a transposable element of the integrative and conjugative element group. Transposable element regulation by antisense transcripts was one of the earliest discovered examples of riboregulation[71], though it has not previously been observed at the scale implied by our RNA-seq analysis. Such antisense regulation could arise spontaneously through capture of transcriptional noise, providing a parsimonious alternative to transcriptional control[72]. It is unclear whether these untranslated transcripts have some function in Ot, or whether they are purely selfish DNA elements that Ot has been unable to dispose of due to its small population size. One intriguing possibility

is that this regulatory mechanism would provide a large pool of double stranded RNA upon intracellular

bacterial lysis, which may explain Ot induction of TLR3 and an antiviral immune response.

In the absence of genetic tools it is difficult to identify specific genes that drive virulence differences

between UT176 and Karp. However, comparative genomics has revealed that whilst the pan genome of

Ot is open, it is largely composed of gene duplications rather than newly acquired genes. This lack of gene

novelty likely reflects the environmental isolation associated with an obligate intracellular lifestyle.

Consequently, strain-specific differences in virulence are likely to be driven largely by differences in

relative gene expression rather than the presence or absence of virulence genes. Consistent with this, we

observed an up-regulation of virulence-associated surface proteins in Karp compared with UT176.

The inflammatory response triggered by Ot infection is a key driver of virulence in scrub typhus. We

compared the response of endothelial cells to the two strains of Ot and found that differential activation

of the immune response correlated with differential outcomes in a scrub typhus mouse model. Whilst both

Karp and UT176 induced an antiviral proinflammatory response, as shown previously[13–17], UT176

strongly induced an IL6-mediated pro-inflammatory response whilst Karp induced an IL33-NOS3-FAS

response, differences likely to influence the relative virulence of these strains.

IL33 was one of the most strongly differentially regulated genes between UT176 and Karp infections (5.1

logFC higher in Karp-infected HUVECs). IL33 has previously been shown to play a role in pathogenesis

in a scrub typhus murine model, using the Karp strain, where it was shown that IL33 levels were increased

during Ot infection, that *IL33*-/- mice showed less severe disease symptoms, and that addition of rIL33

increased severity and mortality[68]. Our observations of reduced induction of *IL33* by the less virulent

UT176 strain further support a role for this cytokine in the pathogenesis of scrub typhus.

In summary, we have used dual RNA-seq to gain insights into the transcriptome structure and mechanisms

of gene regulation in the neglected intracellular pathogen Ot during infection. We provide evidence for

widespread antisense regulation, in particular for the RAGE genes. We identified a relationship between

the relative induction of IL33- and IL6-based gene networks in the host and disease severity. These

findings will lay the groundwork for subsequent studies on the regulation of gene expression in Ot and

mechanisms of pathogenesis. More generally, the present study may serve as a blueprint for the

characterization of further obligate intracellular, genetically intractable bacterial pathogens.

Figure legends

Figure 1. Overview. A. Schematic experimental overview. HUVEC = human umbilical vein endothelial cell. **B.** Growth curve showing replication of Ot in cultured HUVEC cells. Bacteria were grown in 24 well plates and the total bacteria per well is shown. Mean and SD from three independent replicates are shown. **C.** Confocal microscopy images of Ot in HUVEC cells 5 days post infection. Additional images and time points are shown in Supp. Fig. 1 and 2. Blue = DAPI (DNA), Red = Evans blue (host cells), green = Ot labelled with Alexa488-click-methionine. **D.** RNA mapping statistics showing the fraction of host and Ot RNA for each condition. The first replicate of the experiment is shown. Individual results for each replicate are shown in Supp. Fig. 3. **E.** Percentage of RNA-seq reads assigned to different classes of RNA in Karp, UT176 and HUVEC.

Figure 2. RNA biology in Ot. A. Northern blot analysis of core non-coding RNAs in Ot. **B.** Structure of the two-piece tmRNA observed in the Ot transcriptome **C.** RNA-seq read coverage over the tmRNA gene mirrors cleavage observed by Northern blot. **D.** A comparison of genomic synteny of two species within the enterobacteriaceae (*Escherichia coli* MG1655 and *Salmonella enterica* serovar Typhimurium SL1344, top), with synteny between the two Orientia strains from this study (bottom). **E.** Pie charts illustrating the relative abundance of RAGE genes in conserved (top) and strain-specific (bottom) operons. **F.** Visualization of the largest conserved operon in Ot, encoding multiple ribosomal genes, showing RNA-seq coverage in both strains.

Figure 3. Antisense transcription is enriched on mobile genetic elements and is predictive of an absence of protein expression. A. B. Plot showing the relationship between protein expression, defined by LFQs, and transcript expression, defined by TPMs. Genes cluster into two groups based on their protein

expression. The red line indicates the threshold for expressed genes (TPM value equal to 10). **C.** Sense

transcription and the ratio of reads assigned to the antisense and sense strands, showing classification

based on proteomics detection. The red line indicates the sense-antisense ratio (1.059182) above which

translation was not detected by mass spectrometry. **D.** ROC curves evaluating the performance of logistic

regression models to predict protein expression from RNA-seq read counts. Model 1 strictly uses sense

expression, Model 2 the antisense-sense ratio, and Model 3 uses both. Incorporating antisense expression

clearly improves model performance. **E.** Fraction of core genes and RAGE genes in the set of genes with

high antisense-sense ratios, compared to all expressed genes.

Figure 4. Differential bacterial gene expression. A. Heatmap illustrating pathways enriched in

differentially expressed genes. All illustrated categories are more highly expressed in Karp. **B.** Volcano

plot showing the differential expression of bacterial genes in Karp and UT176. Bacterial surface genes

(red) and ankyrin repeat-containing effector proteins (blue) with log fold change ≥ 1 are highlighted.

Figure 5. Ot induces an antiviral interferon response in HUVECs. A. Summary of the host response

showing joint and strain specific responses. The joint response is defined as genes with a logFC > 2 and

FDR-corrected p-value < 0.01 for infection with both Karp and UT176. Strain-specific responses are genes

with a logFC > 2 and FDR-corrected p-value < 0.01 for infection with either Karp or UT176, excluding

genes already included in the joint response. **B.** Activation of multiple genes in the canonical interferon

signaling pathway in Karp infected HUVECs compared with uninfected HUVEC cells. **C.** Heat map

showing up-regulation of genes in the interferon signaling pathway in HUVEC cells infected with Karp

and UT176 compared with uninfected cells.

Figure 6. Karp and UT176 lead to the up-regulation of distinct networks in HUVECs. A. Up-regulation of multiple proinflammatory chemokines and cytokines in HUVECs infected with UT176. **B.** Induction of the IL33-FAS-mediated anoikis network in Karp infected HUVECs.

Figure 7. Karp is more virulent than UT176 in a mouse infection model. A. Weight change over 12 days of infection. **B.** Clinical observation score of mice 12 days post-infection. This number is a composite score based on appetite, activity, and hair coat with higher numbers representing low appetite, low activity, and ruffled fur. Details provided in Supp. Fig. 20. **C.** Bacterial copy number in 100 μl blood taken from euthanized mice 12 days post-infection, measured by qPCR. **D.** The ratio of bacterial DNA copy number to mouse DNA copy number in lung, liver, spleen, and kidney of euthanized mice 12 days post-infection, measured by qPCR. **E.** Lesion scores of Hematoxylin and Eosin stained lung, liver, spleen, and kidneys of euthanized mice 12 days post infection. Scores range from 0 to 5 with 0 representing normal tissue and 5 representing severe lesion damage. All graphs show mean and standard deviation. Statistical significance is calculated using unpaired Student t-test in GraphPad Prism software. $**p \leq 0.01$ $***p \leq 0.001$ $****p \leq 0.0001$ **F.** Images of Hematoxylin and Eosin stained lung tissue of mice infected with buffer, UT176 or Karp. Scale bars = 50 μm. * indicates airway and ** indicates blood vessel. Uninfected control: airway, blood vessel and alveoli all appear normal. UT176-infected lungs: There are diffuse thickening and infiltration of alveolar septa with a mixed population of macrophages and lymphocytes (arrows). There is also mild perivascular lymphohistiocytic inflammation (open arrow). Karp-infected lungs: There is diffuse moderate thickening and infiltration of alveolar septa with a mixed population of macrophages and lymphocytes. The airway (*) is unaffected and normal. Additional figures are shown in Supp. Fig. 20.

Materials and Methods

Growth of Ot and isolation of RNA

The clinical isolate strains (Karp and UT176) of *Orientia tsutsugamushi* were propagated in a confluent

monolayer of host cells (HUVEC, Human Umbilical Vein Endothelial Cells; ATCC PCS-100-010) for 5

days at MOI 100:1. Cells were cultured using Media200 (Thermo Fisher, Catalog no. M200-500)

supplemented with LVES media (Thermo Fisher, Catalog no. A14608-01) at 35 °C and 5% CO_2. The

infectivity was determined by qPCR of the single copy Ot gene *47kDa* at day 5-7. Both uninfected cells

and infected cells were harvested by incubating the cells on ice and quickly resuspending in RNAprotect

Bacteria Reagent (Qiagen, catalog no. 76506), then storing at -80 °C until use. RNA extraction was

performed using the Qiagen RNeasy Plus kit (Qiagen, catalog number 74136) according to manufacturer's

instructions and as described previously (Atwal, 2017).

RNA processing and sequencing

The integrity of the DNase-treated RNA samples was assessed in a Bioanalyzer (Agilent). All samples

had RIN (RNA integrity number) values ≥8.0. Ribosomal transcripts were removed using the Ribo-Zero

Gold (epidemiology) kit (Illumina). Following the manufacturer's instructions, 500 ng of total, DNase-

treated RNA was used as an input to the ribo-depletion procedure. rRNA-depleted RNA was precipitated

in ethanol for 3 h at -20°C.

cDNA libraries for Illumina sequencing were generated by Vertis Biotechnologie AG, Freising-

Weihenstephan, Germany. rRNA-free RNA samples were first sheared via ultrasound sonication (four 30-

s pulses at 4°C) to generate on average 200- to 400-nt fragments. Fragments of 20 nt were removed using

the Agencourt RNAClean XP kit (Beckman Coulter Genomics) and the Illumina TruSeq adapter was

ligated to the 3' ends of the remaining fragments. First-strand cDNA synthesis was performed using M-

MLV reverse transcriptase (NEB) wherein the 3' adapter served as a primer. The first-strand cDNA was

purified, and the 5' Illumina TruSeq sequencing adapter was ligated to the 3' end of the antisense cDNA.

The resulting cDNA was PCR-amplified to about 10 to 20 ng/µl using a high fidelity DNA polymerase.

The TruSeq barcode sequences were part of the 5' and 3' TruSeq sequencing adapters. The cDNA library

was purified using the Agencourt AMPure XP kit (Beckman Coulter Genomics) and analyzed by capillary

electrophoresis (Shimadzu MultiNA microchip).

For sequencing, cDNA libraries were pooled in approximately equimolar amounts. The cDNA pool was

size fractionated in the size range of 200 to 600 bp using a differential cleanup with the Agencourt AMPure

kit (Beckman Coulter Genomics). Aliquots of the cDNA pools were analyzed by capillary electrophoresis

(Shimadzu MultiNA microchip). Sequencing was performed on a NextSeq 500 platform (Illumina) at

Vertis Biotechnologie AG, Freising-Weihenstephan, Germany (single-end mode; 75 cycles).

Northern blots

Each 15 µg of total RNA (i.e. a mixture of human and Ot RNA) prepared as above were loaded per lane

and separated in 6% (vol/vol) polyacrylamide–7 M urea gels. Blotting was performed as previously

described[24]. After the transfer onto Hybond XL membranes (Amersham), RNA was cross-linked with UV

light and hybridized at 42°C with gene-specific 32P-end-labeled DNA oligonucleotides (Supp. Fig. 13)

in Hybri-Quick buffer (Carl Roth AG). After exposure, the screens were read out on a Typhoon FLA 7000

phosphorimager (GE Healthcare).

qRT-PCR

qRT-PCR was performed with the Power SYBR Green RNA-to-CT1-Step kit (Applied Biosystems)

according to the manufacturer's instructions and a CFX96 Touch real-time PCR detection system (Bio-

Rad). Human U6 snRNA served as reference transcripts. Fold changes in expression were determined

using the $2^{(-\Delta\Delta Ct)}$ method [76]. Primer sequences are given in Supp. Fig. 13, and their specificity had been

confirmed using Primer-BLAST (NCBI).

RNA seq read processing and quantification

The raw reads were initially processed according to our established dual RNA-seq pipeline[24]. Briefly, raw

reads were trimmed for adaptor sequences and a minimum read quality of 20 using cutadapt[77]. Reads were

then mapped against the human (GRCh38) and Ot (UT176 accession: LS398547.1; Karp accession:

LS398548.1) reference sequences using the READemption pipeline (v0.4.3, [78]) and segemehl with the

lack remapper (v0.2.0 [79]), removing reads that mapped equally well to the bacterial and host genomes. For

downstream analysis of human gene expression, only uniquely mapping reads were retained for

quantification.

To improve quantification of repetitive sequences, reads mapped to the Ot genomes were used for

quantification of bacterial transcript expression using Salmon (v0.9.1) [29]. Salmon is a quasi-mapping

based gene expression quantification tool that consists of two steps, indexing and quantification.

Transcript fasta files were created from the Genbank annotations using the gene coordinates. The indexing

step was performed in quasi-mapping mode (--type quasi). Expression of the transcripts was quantified

using both stranded forward library type (-lSF) and removing incompatible mappings (--incompatPrior

0.0). Salmon identified identical gene repeats that are collected in 218 groups (see supplementary table,

Karp groups of duplicates). For quantification purposes we retained a single gene from each group.

For the purposes of summarizing gene expression, we calculated mean TPM values from three replicates

for each strain. Genes with a mean TPM greater than 10 were classified as expressed, and those with a

mean TPM value greater than 50 highly expressed.

Gene annotation

For each gene we retrieved the gene name, gene product, and amino acid sequence from the Genbank

annotation. In addition, using eggNOG-mapper[10] we predicted gene names and both KEGG pathways [11]

and GO terms. We manually identified surface antigen encoding proteins using BLAST. The KEGGREST

(Tenenbaum, D (2019) KEGGREST: Client-side REST access to KEGG. R package version 1.18.1.) and

GO.db (Carlson M (2019). GO.db: A set of annotation maps describing the entire Gene Ontology. R

package version 3.5.0.) R packages were used to retrieve KEGG and GO terms, respectively.

Non-coding RNA prediction

Noncoding RNAs were annotated using Rockhopper[76], ANNOgesic [77] (v0.7.17) and Infernal[78] (v1.1.2)

searching sequences against the Rfam database[79]. These provided inconsistent predictions of intergenic

sRNAs. Intergenic sRNAs were manually curated by visual comparison of the predicted sRNA

coordinates with the read coverage in the Integrative Genomics Viewer [80] (v2.5.2). Infernal predicted the

core housekeeping ncRNAs tmRNA, RNaseP, SRP and 5S rRNA. The quantification of the bacterial

transcriptomes complemented with predicted sRNAs was performed using Salmon.

Genomic alignment

540 Genomic comparisons in Figure panels D and F were performed using Easyfig[81]. Escherichia coli K-12

541 MG1655 (Accession number U00096) and Salmonella enterica serovar Typhimurium SL1344 (Accession

542 number FQ312003) were used as comparators for synteny analysis.

Orthology and conserved operon prediction

We predicted orthologous genes between the two Orientia strains using Poff[82] with default parameters in synteny mode. To identify conserved operons we used operon structures predicted in each strain by Rockhopper [41]. Based on visual analysis of read coverage in the Integrative Genomics Viewer, some of the operons were manually extended by addition of genes or merging two operons into one. We also identified partially conserved operons missing some genes in one strain.

Differential gene expression

For the bacteria, differential gene expression analysis was performed between orthologous genes identified by Poff. Genes that were predicted as an orthologous group (more than two genes) were removed from the analysis. Additionally, we removed duplicates (transcripts with perfectly identical sequence) that were identified by Salmon in either strain. For both human and bacterial RNA-seq data, we performed differential gene expression analysis with the edgeR package [83] (v3.20.9) using robust quasi-likelihood estimation [84], including genes with CPM (Counts Per Million) > 10 (for Ot) or CPM > 1 (for HUVEC) in at least 3 libraries. To identify biological processes that differ between two Orientia strains, we have performed gene set analysis using KEGG and GO terms that contain at least 4 expressed genes using the fry test in the edgeR package.

Proteomic sample preparation

Bacteria were propagated in HUVEC cell line at MOI 100:1 and harvested at 5 dpi. Ot was isolated, washed with 0.3 M sucrose, and lysed with 1% triton-X prior to acetone precipitation of protein. Total

protein was then alkylated, reduced, and subsequently treated with Lys-C/Trypsin. Digested peptides were

desalted using Oasis® HLB reversed-phase cartridges, vacuum dried, and stored for MS runs.

Mass spectrometry

The dried samples were resuspended of 2% (v/v) acetonitrile solution containing 0.06% (v/v)

trifluoroacetic acid and 0.5% (v/v) acetic acid and loaded onto an autosampler plate. Online

chromatography was performed using EASY-nLC 1000 (Thermo Scientific) in single-column setup using

0.1% formic acid in water and 0.1% formic acid in acetonitrile as mobile phases. using reversed-phase

C18 column (EASY-Spray LC Column, 75 μm inner diameter x 50 cm, 2 μm particle size) (Thermo

Scientific). The samples were injected and separated on the analytical column maintained at 50 °C using

a 2–23% (v/v) acetonitrile gradient over 60 min, then ramped to 50% over the next 20 min, and finally to

90% within 5 min. The final mixture was maintained for 5 min to elute all remaining peptides. Total run

duration for each sample was 90 min at a constant flow rate of 300 nl/min.

Data was acquired using an Orbitrap Fusion mass spectrometer (Thermo Scientific) in data-dependent

mode. Samples were ionized 2.5 kV and 300 °C at the nanospray source and positively-charged precursor

MS1 signals were detected using an Orbitrap analyzer set to 60,000 resolution, automatic gain control

(AGC) target of 400,000 ions, and maximum injection time (IT) of 50 ms. Precursors with charges 2–7

and having the highest ion counts in each MS1 scan were further fragmented using collision-induced

dissociation (CID) at 35% normalized collision energy and their MS2 signals were analyzed by ion trap

at an AGC of 10,000 and maximum IT of 35 ms. Precursors used for MS2 scans were excluded for 90 s

in order to avoid re-sampling of high abundance peptides. The MS1-MS2 cycles were repeated every 3 s

until completion of the run.

Identification of proteins within each sample was performed using MaxQuant (v1.5.5.1). Raw mass

spectra were searched against Orientia tsutsugamushi primary protein sequences derived from complete

genome data. Human whole proteome sequences were obtained from Uniprot and included as background.

Carbamidomethylation on Cys was set as the fixed modification and acetylation on protein N-terminus

and oxidation of Met were set as dynamic modifications for the search. Trypsin was set as the digestion

enzyme and was allowed up to 3 missed cleavage sites. Precursors and fragments were accepted if they

had a mass error within 20 ppm. Peptides were matched to spectra at a false discovery rate (FDR) of 1%

against the decoy database.

Proteomic data analysis

Protein expression was measured by label free quantification values (LFQs). A protein was classified as

detected if at least two peptides were detected in at least 2 biological replicates, and the mean LFQ across

the three replicates was used for further analysis. Otherwise the protein was classified as undetected, and

the LFQ value was set to zero. The proteomic data includes 19 protein groups that couldn't be resolved,

consisting of 93 proteins. In our analysis we discarded these proteins to simplify the analysis.

Transcript classification

Sense transcript expression was defined by mean TPM value across replicates. The antisense/sense ratio

was calculated as the ratio of mean read counts assigned to the antisense and sense strand of coding

annotations. The duplicated sequences identified by Salmon (Supp. Table 1. Karp groups of duplicates)

and non-coding RNAs were removed from the analysis.

We divided the data set into two classes, detected and undetected in proteomics. Within our analyzed data

set, 322 genes were detected, whereas 1608 genes were not detected by mass spectrometry. We found a

weak positive correlation between TPMs and LFQs for genes with detected proteins (Spearman's

correlation coefficient equal to 0.32), but it was not a linear association (Pearson's correlation coefficient

equal to 0.04). For the further analysis we selected transcripts with sense expression greater than 10 TPMs,

previously defined as our expression threshold.

Logistic Regression model

To test whether antisense-sense ratios are predictive of protein expression, we have applied logistic

regression, which models the probability of a binary response, that is, whether a protein is expressed or

not. We have built 3 competing models. Model 1 makes predictions of the protein expression based solely

on sense transcription:

$\beta_0 + \beta_1 *$ (TPM sense)

Model 2 makes predictions solely on the antisense-sense ratio:

$\beta_0 + \beta_1 *$ ((number of antisense reads) / (number of sense reads))

Model 3 uses both sense transcription and the antisense-sense ratio to make predictions:

$\beta_0 + \beta_1 *$ (TPM sense) $+ \beta_2 *$ ((number of antisense reads) / (number of sense reads))

Since the data is highly imbalanced, 295 transcripts with detected proteins, and 814 without, we used a

downsampling procedure (downSample function) implemented in the caret R package[85] to create a

balanced data set for model training purposes. Next, the function glm() with a logit link function from the

caret package was used to fit models to the reduced data set. For a first indication as to whether any of

these models are predictive, we trained all three models on a downsampled data set consisting of 590

genes, then tested them on the complete data set. To more rigorously assess this result, we have applied

500-fold cross validation. For each fold the data was split randomly into 2 data sets, training and testing

which included 1055 and 54 genes, respectively. Each time the new training data set was reduced to 562

genes, which were used to estimate the model parameters, and then the model was evaluated on the testing

data set. The model performance was evaluated using a variety of measures, i.e. sensitivity, specificity,

accuracy (caret R package) as well as with ROC curves [86] (pROC 1.14.0) and the area under the ROC

curve (AUC).

Immunofluorescence microscopy

The protocol for L-Homopropargylglycine (HPG) incorporation, click chemistry and fluorescence

detection were based on recommendations from Click-iT® HPG Alexa Fluor® Protein Synthesis Assay

Kits (Molecular probe by Life Technologies). HUVECs were grown on chambered coverslip slides (Ibidi,

USA), for 2 days before infection with bacteria at MOI 100:1. To incorporate HPG at times indicated,

medium was removed and replaced with L-methionine-free medium (Dulbecco's Modified Eagle

Medium, DMEM, Cat. no. 21013) containing 25 µM HPG for 30 min at 37 ºC. Labeled bacteria were

washed twice in 1X PBS+ 1mg/ml BSA, pH 7.4 before fixing with 4% formaldehyde and subsequently

being permeabilized with 0.5% TritonX for 20 min on ice. After washing with PBS + 1 mg/ml BSA, the

Click-iT® reaction cocktail (Click-iT® HPG Alexa Fluor® Protein Synthesis Assay Kits cat. C10428)

was incubated with cells for 30 min at room temperature in the dark. The Azide dye (Alexa Fluor®488) was used at a final concentration of 5 μM. After the click reaction, cells were labeled with the actin probe Alexa Fluor® 594 phalloidin at a dilution of 1:40 and the nuclear stain Hoechst diluted to 1:1000 for 30 min at 37 °C. Cells were washed 3X with PBS which was replaced with mounting media after the final wash. Imaging was performed using a Zeiss LSM 7000 equipped with a 63 × 1.4 NA objective lens (Carl Zeiss, USA) and also a Leica SP8 laser scanning confocal microscope.

Analysis of codon bias

We calculated the RSCU (relative synonymous codon usage) for each codon to quantify genome-wide or gene-specific codon usage bias following[52]. To determine the genomic codon counts for each species and gene set, we parsed nucleotide sequence data and annotation in the GenBank file format, downloaded from the NCBI database. We also obtained tRNA gene copy numbers from the GtRNAdb database[87,88], and integrated protein abundance for E. coli K-12 MG1655 data from PaxDB 89.

Host Network/Pathway analysis

To identify pathways that are affected in Karp and/or UT176 infected host cells, genes differentially expressed with an adjusted p-value of < 0.05 were analyzed using Ingenuity Pathway Analysis (IPA) software (Ingenuity® Systems, Inc. Redwood City, CA)[81] as described previously[82]. Selected pathways were chosen based on enrichment p-values and activation Z-scores, and served as the basis for Figs 5, 6, and Supp. Fig. 11, 14, 15, 16, and 17.

Mice and Ethics statement

All animal research was performed strictly under approved Institutional Animal Care and Use Committee

(IACUC) protocol by the IACUC and Biosafety Review Committee at the Armed Forces Research

Institute of Medical Sciences (AFRIMS) Bangkok, Thailand, an AAALAC International-accredited

facility. The protocol number was PN16-05. The animal research was conducted in compliance with Thai

laws, the Animal Welfare Act, and all applicable U.S. Department of Agriculture, Office of Laboratory

Animal Welfare and U.S. Department of Defense guidelines. All animal research adhered to the Guide for

the Care and Use of Laboratory Animals, NRC Publication (8th Edition).

C57BL/6NJcl mice were purchased from Nomura Siam International, Bangkok, Thailand. Mice were

housed in an animal biosafety level 2 facility and moved to an animal biosafety level 3 containment 2 days

before the inoculation. Mice at 6- 8 weeks of age were used in these experiments. Two group of mice (n

=8 per group) were intravenously injected in the tail vein with $1.25x10^6$ genome copies of *O.*

tsutsugamushi of either Karp strain or UT176 strain. The *O. tsutsugamushi* inoculum was derived from

O. tsutsugamushi-infected L929 cells. Clinical signs and body weight were evaluated daily. After 12 days

post inoculation, all mice were euthanized. Blood and tissue samples including lungs, liver, spleen, and

kidneys were collected for bacteria quantification and histopathology.

694
695

Acknowledgements

JS is funded by a Royal Society Dorothy Hodgkin Research Fellowship (DH140154) and this project was

additionally funded by a grant from the University of Oxford Medical Sciences Division Medical Research

Fund. The authors are grateful to Guy Riddihough from Life Science Editors for editorial support on this

manuscript. The authors would like to thank Sandy Pernitzsch/Scigraphix for assistance with Figures 5

and 6.

Author contributions

705 Conceptualization, J.V., L.B., J.S.; Investigation, B.M.G., S.G., A.J.W., J.W., S. C., L.C.W., P.S., L.B.;

706 Formal Analysis, B.M.G., W.K.C., S.S., R.S., L.B.; Resources, R.S., J.V., L.B., J.S.; Writing – Original

707 Draft, L.B., J.S.; Writing – Review and Editing, A.J.W., L.B., J.S.; Supervision, L.B., J.S.; Funding

708 Acquisition, J.V., J.S.

Competing interests

711 The authors declare no competing interests.

References

References
1.   Luce-Fedrow, A., Lehman, M., Kelly, D., Mullins, K., Maina, A., Stewart, R., Ge, H., John, H., Jiang, J. & Richards, A. A Review of Scrub Typhus (Orientia tsutsugamushi and Related Organisms): Then, Now, and Tomorrow. *Tropical Medicine and Infectious Disease* **3**, 8 (2018).
2.   Izzard, L., Fuller, A., Blacksell, S. D., Paris, D. H., Richards, A. L., Aukkanit, N., Nguyen, C., Jiang, J., Fenwick, S., Day, N. P., Graves, S. & Stenos, J. Isolation of a novel Orientia species (O. chuto sp. nov.) from a patient infected in Dubai. *J Clin Microbiol* **48**, 4404-4409 (2010).
3.   Weitzel, T., Dittrich, S., López, J., Phuklia, W., Martinez-Valdebenito, C., Velásquez, K., Blacksell, S. D., Paris, D. H. & Abarca, K. Endemic Scrub Typhus in South America. *N Engl J Med* **375**, 954-961 (2016).
4.   Paris, D. H., Phetsouvanh, R., Tanganuchitcharnchai, A., Jones, M., Jenjaroen, K., Vongsouvath, M., Ferguson, D. P., Blacksell, S. D., Newton, P. N., Day, N. P. & Turner, G. D. Orientia tsutsugamushi in human scrub typhus eschars shows tropism for dendritic cells and monocytes rather than endothelium. *PLoS Negl Trop Dis* **6**, e1466 (2012).
5.   Moron, C. G., Popov, V. L., Feng, H. M., Wear, D. & Walker, D. H. Identification of the target cells of Orientia tsutsugamushi in human cases of scrub typhus. *Mod Pathol* **14**, 752-759 (2001).
6.   Cho, N. H., Kim, H. R., Lee, J. H. & Kim, I. S. The Orientia tsutsugamushi genome reveals massive proliferation of conjugative type IV secretion system and host– cell interaction genes. *PNAS* **104**, 7981-7986 (2007).
7.   Nakayama, K., Yamashita, A., Kurokawa, K., Morimoto, T., Ogawa, M., Fukuhara, M., Urakami, H., Ohnishi, M., Uchiyama, I., Ogura, Y., Ooka, T., Oshima, K., Tamura, A., Hattori, M. & Hayashi, T. The Whole-genome sequencing of the obligate intracellular bacterium Orientia tsutsugamushi revealed massive gene amplification during reductive genome evolution. *DNA Res* **15**, 185-199 (2008).
8.   Batty EM, C. S., Blacksell SB, Richards A, Paris D, Bowden R, Chan C, Lachumanan R, Day N, Donnelly P, Chen SL, Salje J. Long-read whole genome sequencing and comparative analysis of six strains of the human pathogen *Orientia tsutsugamushi*. *Plos Negl Trop Dis* (2018).
9.   Kelly, D. J., Fuerst, P. A., Ching, W. M. & Richards, A. L. Scrub typhus: the geographic distribution of phenotypic and genotypic variants of Orientia tsutsugamushi. *Clin Infect Dis* **48 Suppl 3**, S203-30 (2009).
10.   Premaratna, R., Blanton, L. S., Samaraweera, D. N., de Silva, G. N., Chandrasena, N. T., Walker, D. H. & de Silva, H. J. Genotypic characterization of Orientia tsutsugamushi from patients in two geographical locations in Sri Lanka. *BMC Infect Dis* **17**, 67 (2017).
11.   Varghese, G. M., Janardhanan, J., Mahajan, S. K., Tariang, D., Trowbridge, P., Prakash, J. A., David, T., Sathendra, S. & Abraham, O. C. Molecular Epidemiology and Genetic Diversity of Orientia tsutsugamushi from Patients with Scrub Typhus in 3 Regions of India. *Emerg Infect Dis* **21**, 64-69 (2015).
12.   James, S. L., Blacksell, S. D., Nawtaisong, P., Tanganuchitcharnchai, A., Smith, D. J., Day, N. P. & Paris, D. H. Antigenic Relationships among Human Pathogenic Orientia tsutsugamushi Isolates from Thailand. *PLoS Negl Trop Dis* **10**, e0004723 (2016).
13.   Cho, N., others & Kim, I. S. Expression of Chemokine Genes in Murine Macrophages Infected with Orientia tsutsugamushi. *infection and Immunity* (2000).
14.   Cho, N. H., Seong, S. Y., Choi, M. S. & Kim, I. S. Expression of chemokine genes in human dermal microvascular endothelial cell lines infected with Orientia tsutsugamushi. *Infect Immun* **69**, 1265-1272 (2001).

15.   Koh, Y. S., Yun, J. H., Seong, S. Y., Choi, M. S. & Kim, I. S. Chemokine and cytokine production
during Orientia tsutsugamushi infection in mice. *Microbial pathogenesis* (2004).

16.   Tantibhedhyangkul, W., Prachason, T., Waywa, D., El Filali, A., Ghigo, E., Thongnoppakhun, W.,
Raoult, D., Suputtamongkol, Y., Capo, C., Limwongse, C. & Mege, J. L. Orientia tsutsugamushi
stimulates an original gene expression program in monocytes: relationship with gene expression in
patients with scrub typhus. *PLoS Negl Trop Dis* **5**, e1028 (2011).

17.   Tantibhedhyangkul, W., Ben Amara, A., Textoris, J., Gorvel, L., Ghigo, E., Capo, C. & Mege, J.
L. Orientia tsutsugamushi, the causative agent of scrub typhus, induces an inflammatory program
in human macrophages. *Microb Pathog* **55**, 55-63 (2013).

18.   Astrup, E., Janardhanan, J., Otterdal, K., Ueland, T., Prakash, J. A., Lekva, T., Strand, O. A.,
Abraham, O. C., Thomas, K., Damas, J. K., Mathews, P., Mathai, D., Aukrust, P. & Varghese, G.
M. Cytokine network in scrub typhus: high levels of interleukin-8 are associated with disease
severity and mortality. *PLoS Negl Trop Dis* **8**, e2648 (2014).

19.   Fukuhara, M., Fukazawa, M., Tamura, A., Nakamura, T. & Urakami, H. Survival of two Orientia
tsutsugamushi bacterial strains that infect mouse macrophages with varying degrees of virulence.
Microb Pathog **39**, 177-187 (2005).

20.   Hanson, B. Comparative susceptibility to mouse interferons of Rickettsia tsutsugamushi strains
with different virulence in mice and of Rickettsia rickettsii. *Infect Immun* **59**, 4134-4141 (1991).

21.   Sunyakumthorn, P., Paris, D. H., Chan, T. C., Jones, M., Luce-Fedrow, A., Chattopadhyay, S.,
Jiang, J., Anantatat, T., Turner, G. D., Day, N. P. & Richards, A. L. An intradermal inoculation
model of scrub typhus in Swiss CD-1 mice demonstrates more rapid dissemination of virulent
strains of Orientia tsutsugamushi. *PLoS One* **8**, e54570 (2013).

22.   Groves, M. G. & Kelly, D. J. Characterization of factors determining Rickettsia tsutsugamushi
pathogenicity for mice. *Infect Immun* **57**, 1476-1482 (1989).

23.   Saliba, A. E., C Santos, S. & Vogel, J. New RNA-seq approaches for the study of bacterial
pathogens. *Curr Opin Microbiol* **35**, 78-87 (2017).

24.   Westermann, A. J., Förstner, K. U., Amman, F., Barquist, L., Chao, Y., Schulte, L. N., Müller, L.,
Reinhardt, R., Stadler, P. F. & Vogel, J. Dual RNA-seq unveils noncoding RNA functions in host-
pathogen interactions. *Nature* (2016).

25.   Westermann, A. J., Barquist, L. & Vogel, J. Resolving host-pathogen interactions by dual RNA-
seq. *PLOS Pathog* **13**, (2017).

26.   Humphrys, M. S., Creasy, T., Sun, Y., Shetty, A. C., Chibucos, M. C., Drabek, E. F., Fraser, C.
M., Farooq, U., Sengamalay, N., Ott, S., Shou, H., Bavoil, P. M., Mahurkar, A. & Myers, G. S.
Simultaneous transcriptional profiling of bacteria and their host cells. *PLoS One* **8**, e80597 (2013).

27.   Enatsu, T., Urakami, H. & Tamura, A. Phylogenetic analysis of Orientia tsutsugamushi strains
based on the sequence homologies of 56-kDa type-specific antigen genes. *FEMS Microbiol Lett*
180, 163-169 (1999).

28.   Paris, D. H., Aukkanit, N., Jenjaroen, K., Blacksell, S. D. & Day, N. P. A highly sensitive
quantitative real-time PCR assay based on the groEL gene of contemporary Thai strains of
Orientia tsutsugamushi. *Clin Microbiol Infect* **15**, 488-495 (2009).

29.   Patro, R., Duggal, G., Love, M. I., Irizarry, R. A. & Kingsford, C. Salmon provides fast and bias-
aware quantification of transcript expression. *Nat Methods* **14**, 417-419 (2017).

30.   Withey, J. H. & Friedman, D. I. A salvage pathway for protein structures: tmRNA and trans-
translation. *Annu Rev Microbiol* **57**, 101-123 (2003).

31.   Julio, S. M., Heithoff, D. M. & Mahan, M. J. ssrA (tmRNA) plays a role in Salmonella enterica
serovar Typhimurium pathogenesis. *J Bacteriol* **182**, 1558-1563 (2000).

32.   Svetlanov, A., Puri, N., Mena, P., Koller, A. & Karzai, A. W. Francisella tularensis tmRNA system
807   mutants are vulnerable to stress, avirulent in mice, and provide effective immune protection. *Mol*
808   *Microbiol* **85**, 122-141 (2012).
809   33.   Mao, C., Bhardwaj, K., Sharkady, S. M., Fish, R. I., Driscoll, T., Wower, J., Zwieb, C., Sobral, B.
810   W. & Williams, K. P. Variations on the tmRNA gene. *RNA Biol* **6**, 355-361 (2009).
811   34.   Keiler, K. C., Shapiro, L. & Williams, K. P. tmRNAs that encode proteolysis-inducing tags are
812   found in all known bacterial genomes: A two-piece tmRNA functions in Caulobacter. *Proc Natl*
813   *Acad Sci U S A* **97**, 7778-7783 (2000).
814   35.   Gaudin, C., Zhou, X., Williams, K. P. & Felden, B. Two-piece tmRNA in cyanobacteria and its
815   structural analysis. *Nucleic Acids Res* **30**, 2018-2024 (2002).
816   36.   Sharkady, S. M. & Williams, K. P. A third lineage with two-piece tmRNA. *Nucleic Acids Res* **32**,
817   4531-4538 (2004).
818   37.   Storz, G., Vogel, J. & Wassarman, K. M. Regulation by small RNAs in bacteria: expanding
819   frontiers. *Mol Cell* **43**, 880-891 (2011).
820   38.   Wagner, E. G. H. & Romby, P. Small RNAs in bacteria and archaea: who they are, what they do,
821   and how they do it. *Adv Genet* **90**, 133-208 (2015).
822   39.   Albrecht, M., Sharma, C. M., Dittrich, M. T., Müller, T., Reinhardt, R., Vogel, J. & Rudel, T. The
823   transcriptional landscape of Chlamydia pneumoniae. *Genome Biol* **12**, R98 (2011).
824   40.   Dugar, G., Herbig, A., Förstner, K. U., Heidrich, N., Reinhardt, R., Nieselt, K. & Sharma, C. M.
825   High-resolution transcriptome maps reveal strain-specific regulatory features of multiple
826   Campylobacter jejuni isolates. *PLoS Genet* **9**, e1003495 (2013).
827   41.   Kröger, C., Dillon, S. C., Cameron, A. D., Papenfort, K., Sivasankaran, S. K., Hokamp, K., Chao,
828   Y., Sittka, A., Hébrard, M., Händler, K., Colgan, A., Leekitcharoenphon, P., Langridge, G. C.,
829   Lohan, A. J., Loftus, B., Lucchini, S., Ussery, D. W., Dorman, C. J., Thomson, N. R., Vogel, J. &
830   Hinton, J. C. The transcriptional landscape and small RNAs of Salmonella enterica serovar
831   Typhimurium. *Proc Natl Acad Sci U S A* **109**, E1277-86 (2012).
832   42.   Sharma, C. M., Hoffmann, S., Darfeuille, F., Reignier, J., Findeiss, S., Sittka, A., Chabas, S.,
833   Reiche, K., Hackermüller, J., Reinhardt, R., Stadler, P. F. & Vogel, J. The primary transcriptome
834   of the major human pathogen Helicobacter pylori. *Nature* **464**, 250-255 (2010).
835   43.   Toledo-Arana, A., Dussurget, O., Nikitas, G., Sesto, N., Guet-Revillet, H., Balestrino, D., Loh, E.,
836   Gripenland, J., Tiensuu, T., Vaitkevicius, K., Barthelemy, M., Vergassola, M., Nahori, M. A.,
837   Soubigou, G., Régnault, B., Coppée, J. Y., Lecuit, M., Johansson, J. & Cossart, P. The Listeria
838   transcriptional landscape from saprophytism to virulence. *Nature* **459**, 950-956 (2009).
839   44.   Vogel, J., Bartels, V., Tang, T. H., Churakov, G., Slagter-Jäger, J. G., Hüttenhofer, A. & Wagner,
840   E. G. RNomics in Escherichia coli detects new sRNA species and indicates parallel transcriptional
841   output in bacteria. *Nucleic Acids Res* **31**, 6435-6443 (2003).
842   45.   Tjaden, B. De novo assembly of bacterial transcriptomes from RNA-seq data. *Genome Biol* **16**, 1
843   (2015).
844   46.   Andersson, S. G., Zomorodipour, A., Andersson, J. O., Sicheritz-Ponten, T., Alsmark, U. C.,
845   Podowski, R. M., Naslund, A. K., Eriksson, A. S., Winkler, H. H. & Kurland, C. G. The genome
846   sequence of Rickettsia prowazekii and the origin of mitochondria. *Nature* **396**, 133-140 (1998).
847   47.   Georg, J. & Hess, W. R. Widespread Antisense Transcription in Prokaryotes. *Microbiol Spectr* **6**,
848   (2018).
849   48.   Wade, J. T. & Grainger, D. C. Pervasive transcription: illuminating the dark matter of bacterial
850   transcriptomes. *Nat Rev Microbiol* **12**, 647-653 (2014).

49. Raghavan, R., Sloan, D. B. & Ochman, H. Antisense transcription is pervasive but rarely
conserved in enteric bacteria. *MBio* **3**, (2012).
50. Lloréns-Rico, V., Cano, J., Kamminga, T., Gil, R., Latorre, A., Chen, W. H., Bork, P., Glass, J. I.,
Serrano, L. & Lluch-Senar, M. Bacterial antisense RNAs are mainly the product of transcriptional
noise. *Sci Adv* **2**, e1501363 (2016).
51. Andersson, S. G. & Sharp, P. M. Codon usage and base composition in Rickettsia prowazekii. *J*
Mol Evol **42**, 525-536 (1996).
52. Botzman, M. & Margalit, H. Variation in global codon usage bias among prokaryotic organisms is
associated with their lifestyles. *Genome Biol* **12**, R109 (2011).
53. Sharp, P. M., Bailes, E., Grocock, R. J., Peden, J. F. & Sockett, R. E. Variation in the strength of
selected codon usage bias among bacteria. *Nucleic Acids Res* **33**, 1141-1153 (2005).
54. Duret, L. tRNA gene number and codon usage in the C. elegans genome are co-adapted for
optimal translation of highly expressed genes. *Trends Genet* **16**, 287-289 (2000).
55. Ikemura, T. Correlation between the abundance of Escherichia coli transfer RNAs and the
occurrence of the respective codons in its protein genes: a proposal for a synonymous codon
choice that is optimal for the E. coli translational system. *J Mol Biol* **151**, 389-409 (1981).
56. Ikemura, T. Correlation between the abundance of yeast transfer RNAs and the occurrence of the
respective codons in protein genes. Differences in synonymous codon choice patterns of yeast and
Escherichia coli with reference to the abundance of isoaccepting transfer RNAs. *J Mol Biol* **158**,
573-597 (1982).
57. Plotkin, J. B. & Kudla, G. Synonymous but not the same: the causes and consequences of codon
bias. *Nat Rev Genet* **12**, 32-42 (2011).
58. Frumkin, I., Lajoie, M. J., Gregg, C. J., Hornung, G., Church, G. M. & Pilpel, Y. Codon usage of
highly expressed genes affects proteome-wide translation efficiency. *Proc Natl Acad Sci U S A*
115, E4940-E4949 (2018).
59. Goodman, D. B., Church, G. M. & Kosuri, S. Causes and effects of N-terminal codon bias in
bacterial genes. *Science* **342**, 475-479 (2013).
60. Ha, N. Y., Kim, Y., Choi, J. H., Choi, M. S., Kim, I. S., Kim, Y. S. & Cho, N. H. Detection of
antibodies against Orientia tsutsugamushi Sca proteins in scrub typhus patients and genetic
variation of sca genes of different strains. *Clin Vaccine Immunol* **19**, 1442-1451 (2012).
61. Ha, N. Y., Sharma, P., Kim, G., Kim, Y., Min, C. K., Choi, M. S., Kim, I. S. & Cho, N. H.
Immunization with an Autotransporter Protein of Orientia tsutsugamushi Provides Protective
Immunity against Scrub Typhus. *PLoS Negl Trop Dis* **9**, e0003585 (2015).
62. Ha, N. Y., Shin, H. M., Sharma, P., Cho, H. A., Min, C. K., Kim, H. I., Yen, N. T., Kang, J. S.,
Kim, I. S., Choi, M. S., Kim, Y. K. & Cho, N. H. Generation of protective immunity against
Orientia tsutsugamushi infection by immunization with a zinc oxide nanoparticle combined with
ScaA antigen. *J Nanobiotechnology* **14**, 76 (2016).
63. VieBrock, L., Evans, S. M., Beyer, A. R., Larson, C. L., Beare, P. A., Ge, H., Singh, S., Rodino,
K. G., Heinzen, R. A., Richards, A. L. & Carlyon, J. A. Orientia tsutsugamushi ankyrin repeat-
containing protein family members are Type 1 secretion system substrates that traffic to the host
cell endoplasmic reticulum. *Front Cell Infect Microbiol* **4**, 186 (2014).
64. Evans, S. M., Rodino, K. G., Adcox, H. E. & Carlyon, J. A. Orientia tsutsugamushi uses two Ank
effectors to modulate NF-κB p65 nuclear transport and inhibit NF-κB transcriptional activation.
PLoS Pathog **14**, e1007023 (2018).

65. Cho, K. A., Jun, Y. H., Suh, J. W., Kang, J. S., Choi, H. J. & Woo, S. Y. Orientia tsutsugamushi
induced endothelial cell activation via the NOD1-IL-32 pathway. *Microb Pathog* **49**, 95-104
(2010).
66. Gharaibeh, M., Hagedorn, M., Lilla, S., Hauptmann, M., Heine, H., Fleischer, B. & Keller, C.
Toll-Like Receptor 2 Recognizes Orientia tsutsugamushi and Increases Susceptibility to Murine
Experimental Scrub Typhus. *Infect Immun* **84**, 3379-3387 (2016).
67. Kawai, T. & Akira, S. Signaling to NF-kappaB by Toll-like receptors. *Trends Mol Med* **13**, 460-
469 (2007).
68. Shelite, T. R., Liang, Y., Wang, H., Mendell, N. L., Trent, B. J., Sun, J., Gong, B., Xu, G., Hu, H.,
Bouyer, D. H. & Soong, L. IL-33-Dependent Endothelial Activation Contributes to Apoptosis and
Renal Injury in Orientia tsutsugamushi-Infected Mice. *PLoS Negl Trop Dis* **10**, e0004467 (2016).
69. McLeod, M. P., Qin, X., Karpathy, S. E., Gioia, J., Highlander, S. K., Fox, G. E., McNeill, T. Z.,
Jiang, H., Muzny, D., Jacob, L. S., Hawes, A. C., Sodergren, E., Gill, R., Hume, J., Morgan, M.,
Fan, G., Amin, A. G., Gibbs, R. A., Hong, C., Yu, X. J., Walker, D. H. & Weinstock, G. M.
Complete genome sequence of Rickettsia typhi and comparison with sequences of other
rickettsiae. *J Bacteriol* **186**, 5842-5855 (2004).
70. Yildirim, A. & Feig, M. High-resolution 3D models of Caulobacter crescentus chromosome reveal
genome structural variability and organization. *Nucleic Acids Res* **46**, 3937-3952 (2018).
71. Ellis, M. J. & Haniford, D. B. Riboregulation of bacterial and archaeal transposition. *Wiley*
Interdiscip Rev RNA **7**, 382-398 (2016).
72. Jose, B. R., Gardner, P. P. & Barquist, L. Transcriptional noise and exaptation as sources for
bacterial sRNAs. *Biochem Soc Trans* **47**, 527-539 (2019).
73. Narra, H. P., Schroeder, C. L., Sahni, A., Rojas, M., Khanipov, K., Fofanov, Y. & Sahni, S. K.
Small Regulatory RNAs of Rickettsia conorii. *Sci Rep* **6**, 36728 (2016).
74. Schroeder, C. L., Narra, H. P., Rojas, M., Sahni, A., Patel, J., Khanipov, K., Wood, T. G., Fofanov,
Y. & Sahni, S. K. Bacterial small RNAs in the Genus Rickettsia. *BMC Genomics* **16**, 1075 (2015).
75. Schroeder, C. L., Narra, H. P., Sahni, A., Rojas, M., Khanipov, K., Patel, J., Shah, R., Fofanov, Y.
& Sahni, S. K. Identification and Characterization of Novel Small RNAs in Rickettsia prowazekii.
Front Microbiol **7**, 859 (2016).
76. Livak, K. J. & Schmittgen, T. D. Analysis of relative gene expression data using real-time
quantitative PCR and the 2(-Delta Delta C(T)) Method. *Methods* **25**, 402-408 (2001).
77. Martin, M. Cutadapt Removes Adapter Sequences From High-throughput Sequencing Reads.
EMBnet.journal **17**, 10-12 (2011).
78. Förstner, K. U., Vogel, J. & Sharma, C. M. READemption-a tool for the computational analysis of
deep-sequencing-based transcriptome data. *Bioinformatics* **30**, 3421-3423 (2014).
79. Otto, C., Stadler, P. F. & Hoffmann, S. Lacking alignments? The next-generation sequencing
mapper segemehl revisited. *Bioinformatics* **30**, 1837-1843 (2014).
80. Huerta-Cepas, J., Szklarczyk, D., Forslund, K., Cook, H., Heller, D., Walter, M. C., Rattei, T.,
Mende, D. R., Sunagawa, S., Kuhn, M., Jensen, L. J., von Mering, C. & Bork, P. eggNOG 4.5: a
hierarchical orthology framework with improved functional annotations for eukaryotic,
prokaryotic and viral sequences. *Nucleic Acids Res* **44**, D286-93 (2016).
81. Krämer, A., Green, J., Pollard, J. & Tugendreich, S. Causal analysis approaches in Ingenuity
Pathway Analysis. *Bioinformatics* **30**, 523-530 (2014).
82. Subbian, S., Bandyopadhyay, N., Tsenova, L., O'Brien, P., Khetani, V., Kushner, N. L., Peixoto,
B., Soteropoulos, P., Bader, J. S., Karakousis, P. C., Fallows, D. & Kaplan, G. Early innate

immunity determines outcome of Mycobacterium tuberculosis pulmonary infection in rabbits. *Cell*
Commun Signal **11**, 60 (2013).
942

Figure 1. Overview. A. Schematic experimental overview. HUVEC = human umbilical vein endothelial cell. **B.** Growth curve showing replication of Ot in cultured HUVEC cells. Bacteria were grown in 24 well plates and the total bacteria per well is shown. Mean and SD from three independent replicates are shown. **C.** Confocal microscopy images of Ot in HUVEC cells 5 days post infection. Additional images and time points are shown in Supp. Fig. 1 and 2. Blue = DAPI (DNA), Red = Evans blue (host cells), green = Ot labelled with Alexa488-click-methionine. **D.** RNA mapping statistics showing the fraction of host and Ot RNA for each condition. The mean of three independent experiments is shown. Individual results for each replicate are shown in Supp. Fig. 3. **E.** Percentage of RNA-seq reads assigned to different classes of RNA in Karp, UT176 and HUVEC.

Figure 2. RNA biology in Ot. A. Northern blot analysis of core non-coding RNAs in Ot. **B.** Structure of the two-piece tmRNA observed in the Ot transcriptome **C.** RNA-seq read covereage over the tmRNA gene mirrors cleavage observed by Northern blot. **D.** A comparison of genomic synteny of two species within the enterobacteriaceae (*Escherichia coli* MG1655 and *Salmonella enterica* serovar Typhimurium SL1344, top), with synteny between the two Orientia strains from this study (bottom). **E.** Pie charts illustrating the relative abundance of RAGE genes in conserved (right) and strain-specific (left) operons. **F.** Visualization of the largest conserved operon in Ot, encoding multiple ribosomal genes, showing RNA-seq coverage in both strains.

Figure 3. Antisense transcription is enriched on mobile genetic elements and is predictive of an absence of protein expression.
A. B. Plot showing the relationship between protein expression, defined by LFQs, and transcript expression, defined by TPMs. Genes cluster into two groups based on their protein expression. The red line indicates the threshold for expressed genes (TPM value equal to 10). **C.** Sense transcription and the ratio of reads assigned to the antisense and sense strands, showing classification based on proteomics detection. The red line indicates the sense-antisense ratio (1.059182) above which translation was not detected by mass spectrometry. **D.** ROC curves evaluating the performance of logistic regression models to predict protein expression from RNA-seq read counts. Model 1 strictly uses sense expression, Model 2 the antisense-sense ratio, and Model 3 uses both. Incorporating antisense expression clearly improves model performance. **E.** Fraction of core genes and RAGE genes in the set of genes with high antisense-sense ratios, compared to all expressed genes.

Figure 4. Differential bacterial gene expression. A. Heatmap illustrating pathways enriched in differentially expressed genes. All illustrated categories are more highly expressed in Karp. **B**. Volcano plot showing differential expression of bacterial genes in Karp and UT176. Bacterial surface genes (red) and ankyrin repeat containing effector proteins (blue) with log fold change ≥1 are highlighted.

Figure 5. Ot induces an antiviral interferon response in HUVECs. A. Summary of the host response showing joint and strain specific responses. **B.** Activation of multiple genes in the canonical interferon signaling pathway in Karp infected HUVECs compared with uninfected HUVEC cells. **C.** Heat map showing up-regulation of genes in the interferon signaling pathway in HUVECs infected with Karp and UT176 compared with uninfected cells.

Figure 6. Karp and UT176 lead to up-regulation of distinct networks in HUVECs. A. Up-regulation of multiple proinflammatory chemokines and cytokines in HUVECs infected with UT176. **B.** Induction of the IL33-FAS-mediated anoikis network in Karp infected HUVEC cells.

Figure 7. Karp is more virulent than UT176 in a mouse infection model. A. Weight change over 12 days infection. **B.** Clinical observation score of mice 12 days post infection. This number is a composite score based on appetite, activity, and hair coat with higher numbers representing low appetite, low activity, and ruffled fur. Details provided in Supp. Fig. 20. **C.** Bacterial copy number in 100 μl blood taken from euthanized mice 12 days post infection, measured by qPCR. **D.** Ratio of bacterial DNA copy number to mouse DNA copy number in lung, liver, spleen, and kidney of euthanized mice 12 days post infection, measured by qPCR. **E.** Lesion scores of Hematoxylin and Eosin stained lung, liver, spleen, and kidneys of euthanized mice 12 days post infection. Scores range from 0 to 5 with 0 representing normal tissue and 5 representing severe lesion damage. All graphs show mean and standard deviation. Statistical significance is calculated using unpaired t test using GraphPad Prism software. **p≤0.01 ***p≤0.001 ****p≤0.0001 **F.** Images of Hematoxylin and Eosin stained lung tissue of mice infected with buffer, UT176 or Karp. Scale bars = 50 μm. * indicates airway and ** indicates blood vessel. Uninfected control: airway, blood vessel and alveoli all appear normal. UT176-infected lungs: There is diffuse thickening and infiltration of alveolar septa with a mixed population of macrophages and lymphocytes (arrows). There is also mild perivascular lymphohistiocytic inflammation (open arrow). Karp-infected lungs: There is diffuse moderate thickening and infiltration of alveolar septa with a mixed population of macrophages and lymphocytes. The airway (*) is unaffected and normal. Additional figures are shown in Supp. Fig. 20.