Time-resolved dual RNA-Seq reveals extensive rewiring of lung epithelial and pneumococcal transcriptomes during early infection

Rieza Aprianto[1], Jelle Slager[1], Siger Holsappel[1] and Jan-Willem Veening[1,*]

[1]Molecular Genetics Group, Groningen Biomolecular Sciences and Biotechnology Institute, Centre for Synthetic Biology, University of Groningen, Nijenborgh 7, 9747 AG Groningen, The Netherlands.

*Correspondence to: j.w.veening@rug.nl, Tel: +31 (0)50 3632408, @jwveening

Keywords: *Streptococcus pneumoniae*, transcriptomics, dual RNA-Seq, host-pathogen interaction, adherence, lung epithelial cells

Main Points

- Early pneumococcal infection model for dual RNA-Seq has been established
- Simultaneous RNA isolation, rRNA depletion and sequencing are suitable for dual RNA-Seq
- Gene expression data of host and pathogen is presented in an online user-friendly database (http://dualrnaseq.molgenrug.nl)
- A capsular mutant revealed adherence-specific host and pathogen transcriptional changes: repression of innate epithelial immune response and activation of pneumococcal sugar importers

Graphical Abstract

Abstract

Streptococcus pneumoniae (pneumococcus) is the main etiological agent of pneumonia. Pneumococcal pneumonia is initiated by bacterial adherence to lung epithelial cells. Infection to the epithelium is a disruptive interspecies interaction involving numerous transcription-mediated processes. Revealing transcriptional changes may provide valuable insights into pneumococcal disease. Dual RNA-Seq allows simultaneous monitoring of the transcriptomes of both host and pathogen. Here, we developed a time-resolved infection model of human lung alveolar epithelial cells by *S. pneumoniae* and assessed transcriptome changes by dual RNA-Seq. Our data provide new insights into host-microbe interactions and show that the epithelial glutathione-detoxification pathway is activated by bacterial presence. We observed that adherent pneumococci, not free-floating bacteria, access host-associated carbohydrates and repress innate immune responses. In conclusion, we provide a dynamic dual-transcriptomics overview of early pneumococcal infection with easy online access (http://dualrnaseq.molgenrug.nl). Further database exploration may expand our understanding of epithelial-pneumococcal interaction, leading to novel antimicrobial strategies.

Introduction

Lower respiratory tract infections (LRTIs), or pneumonia, claim more lives than any other communicable disease worldwide; the main etiologic agent behind this infection is the Gram-positive opportunistic pathogen *Streptococcus pneumoniae* (pneumococcus, Prina et al., 2015). Normally part of the human nasopharyngeal microflora, *S. pneumoniae* can invade the lower airways where it provokes host inflammatory and immune responses (Kadioglu et al., 2008). At the earliest stage of infection, pneumococcus adheres to epithelial cells and interacts intimately with the epithelium (Hammerschmidt et al., 2007). Meanwhile, host and pathogen cross-communicate and, they simultaneously, affect each other in a disruptive manner (Kadioglu et al., 2008; Lee et al., 2004). This interspecies interaction activates numerous processes in epithelial and pneumococcal cells (Bootsma et al., 2007; Mlacha et al., 2013). To obtain comprehensive and meaningful biological knowledge of the infection processes involved in pathogenesis, simultaneous monitoring of the transcriptome changes in both species is required (Westermann et al., 2012).

Lung epithelial cells perform vital roles during infection. First, the cells form a physical barrier to the external environment. On top of these cells, a thick layer of epithelial-derived mucus offers extra protection that traps and removes pathogens (Voynow and Rubin, 2009). Mucins, the main component of mucus, are large glycoprotein-polymers rich in sialic acids and other aminosaccharides (Rose and Voynow, 2006). Additionally, epithelial cells kill pathogens directly by producing antimicrobial peptides, e.g.: defensins and cathelicidins (Tecle et al., 2010). Moreover, epithelial cells regulate innate immune responses by secreting a wide array of pro-inflammatory cytokines, that recruit neutrophils and activate macrophages (Hallstrand et al., 2014). Finally, epithelial cells activate adaptive immune cells, including dendritic cells and T-cells via chemokine expression (Soumelis et al., 2002).

Pneumococcal adherence to epithelial cells is the first necessary step to

pathogenesis (Bogaert et al., 2004). In order to adhere, pneumococcus must quickly shed

the thick exopolysaccharide capsule, which protects against phagocytes (Abeyta et al.,

2003; Hyams et al., 2010). The shedding exposes surface adhesion factors and

desensitizes the bacterium from antimicrobial peptides (Beiter et al., 2008; Kietzman et al.,

2016). Subsequently, *S. pneumoniae* must acquire nutrients to support growth and, at the

same time, evade host immune responses (Shelburne et al., 2008). Pneumococcal factors

may be involved in multiple processes, e.g.: PsaA, a surface-exposed protein, acts

concurrently as adhesion factor and manganese transporter (Rajam et al., 2008). The

scarce manganese (Gray et al., 2010) helps in neutralizing reactive oxygen species (ROS)

and bacterial fitness (Tseng et al., 2002).

Interspecies interaction during infection is a chaotic process which necessitates rapid

and massive adaptation for epithelial and pneumococcal survival. During the adaptation,

transcriptional changes plays a focal point both in host (Jenner and Young, 2005) and in

pathogen (Sorek and Cossart, 2010). RNA-sequencing (RNA-Seq) delivers genome-wide

quantitative snapshots of the transcriptome (Kukurba and Montgomery, 2015). In a thought

experiment, Vogel and co-workers argued that simultaneous profiling of host and pathogen

transcriptomics by dual RNA-Seq might provide valuable insights for infection biology

(Westermann et al., 2012). Recent dual RNA-Seq studies were successful in elucidating the

role of sRNAs in the intracellular pathogen *Salmonella typhimurium* (Westermann et al.,

2016), cross-talk in the Gram-negative LRTI pathogen *Haemophilus influenzae* (Baddal et

al., 2015) and transcription profiles in the protozoan *Leishmania major* (Dillon et al., 2015)

during their respective infection.

Here, we exploited the dual RNA-Seq approach to simultaneously monitor the transcriptome cross-talk between lung alveolar epithelial cells and pneumococci during early infection. Due to the transient and highly-dynamic nature of the transcriptome (Pedersen et al., 2011), we monitored the transcriptional changes in a time-resolved manner. Moreover, since pneumococcal adherence to epithelial cells determines the outcome of early infection, we compared adherent and non-adherent *S. pneumoniae* to allow specific transcriptional interrogation on adherence. Additionally, we confirmed our dual RNA-Seq gene expression data by qRT-PCR and quantitative fluorescence microscopy to visualize pneumococcal proteins and thereby confirm several novel biological observations identified in the dataset. Finally, we developed a user-friendly online database (http://dualrnaseq.molgenrug.nl), giving access of our detailed time-resolved dual transcriptomes data to the pneumococcal, microbiology, immunology and pulmonology research communities.

Results

The early pneumococcal infection model to epithelial cells

At the first phase of LRTI, *S. pneumoniae* adheres to the sterile apical-side of epithelial cells, adaptation of bacterial transcriptome occurs followed by bacterial outgrowth (Mlacha et al., 2013), which in turn stimulate epithelial transcriptional responses to their presence (Bootsma et al., 2007). We aimed to recapitulate these events in an *in vitro* model consisting of co-incubation of the pathogenic *S. pneumoniae* strain D39 (serotype 2) to a confluent layer of type II lung alveolar human epithelial cell (A549) at a multiplicity of infection (MOI) of 10, i.e., ten pneumococci per epithelial cell (**Fig.1A**). Five time points up

to 4 hours after infection were selected to capture both early transcriptome responses (30 and 60 mpi) and later responses (120 and 240 mpi). Six technical replicates (individual wells) were pooled into one biological replicate. Two biological replicates were used for each time point, except for 240 mpi where we only obtained one replicate (**Fig.1E**). In order to elucidate adherence-specific expression, we incorporated an isogenic unencapsulated D39 strain (Δ*cps2E*) with increased adherence to epithelial cells into the model (Kjos et al., 2015). The capsular mutant showed significantly greater capacity to adhere to epithelial cells than its encapsulated parental strain ($p<0.001$, **Fig.1C**). During infection, the total number of cells of both strains were significantly ($p<0.01$) increased after 4 hours (**Fig.1D**), showing that both strains multiply in the model, thereby recapitulating one of the characteristics of infection. To minimize transcriptional changes because of sample handling, we did not separate cellular mixtures before total RNA isolation (epithelial cells, adherent pneumococci and free-floating pneumococci, **Fig.1E**, see **Supplementary Information**).

To analyze the time-resolved dual RNA-Seq dataset, a combination of freely available bioinformatics tools was used (**Fig.1F**). First, raw reads were trimmed (Bolger et al., 2014) and aligned (Dobin et al., 2013) to a chimeric genome containing the concatenated genome of *Homo sapiens* (GRCh38, Ensembl) and *S. pneumoniae* D39 (NC_008533.1, NCBI). The one-step mapping was chosen to minimize rates of false negatives. Reads were then separately counted (Liao et al., 2014) and classified as either epithelial or pneumococcal. Following differential gene expression analysis, three groups of genes were removed (see below) and unbiased automatic clustering (Kumar and E. Futschik, 2007) and functional enrichment were performed (Dennis et al., 2003; van Opijnen and Camilli, 2012).

Dual RNA-Seq generates high-quality datasets with clusters of epithelial and

pneumococcal co-expressed genes

Dual RNA-seq generated single end 75 nt reads. We sequenced to such depth, that on

average each library has 70 million reads (30 to 95 million). After adapter trimming and

removal of low-quality reads, we retained 92% (89.0-93.0%). Subsequently, 79% of reads

(74.6 to 83.0%) successfully aligned to the chimeric genome. Additionally, we concluded

that dual rRNA depletion was successful since only 0.03% of human and 0.24% of

pneumococcal reads mapped as ribosomal RNAs. For each library, we counted 18 million

epithelial reads (26%) and 37 million pneumococcal reads (52%, **Fig.2A**). The high number

of reads in each library and the high fraction of usable reads highlights the quality and

suitability of our approach for dual RNA-Seq: simultaneous total RNA isolation, dual rRNA

depletion and cDNA library preparation. PCA analysis showed no evidence for batch effects

(**Supplementary Figure 1**). Relative enrichment of pneumococcal reads may stem from

the total RNA isolation protocol. Nevertheless, each library contained sufficient epithelial

reads for differential gene expression analysis (The ENCODE Consortium, 2011).

To simplify further analyses, we excluded three gene fractions (**Fig.2B**). First, we

removed unexpressed genes, i.e., without any counts in all libraries: 22.991 (38%) epithelial

genes and 24 (1%) pneumococcal genes. The relatively large fraction of unexpressed

epithelial genes is in accordance with recent studies on the human epithelial transcriptome

(Hackett et al., 2012; St-Pierre et al., 2013). Second, we excluded genes that were

differentially expressed ($p<0.05$) at 0 mpi between unencapsulated ($\Delta cps2E$) and

encapsulated (wt) libraries. While only five epithelial genes were removed, 394 (19%)

pneumococcal genes were already differentially expressed at 0 mpi. Although polar effect

due to *cps2E* disruption can explain differential expression of genes in the 17kb long *cps*

operon, it remains unknown why other genes were differentially expressed. We speculate that constructing the thick exopolysaccharide capsule requires specific transcriptional fine-tuning of numerous genes outside the *cps* locus. Finally, we removed genes with no significant difference ($p>0.05$) and genes with fold changes less than 2 in all contrasts (**Supplementary Figure 2**). In total, the epithelial working libraries contained 4.009 (7% of total) genes and pneumococcal working libraries 868 (42% of total) genes.

To compare gene expression, we normalized expression values using DESeq2 (Love et al., 2014), centered and clustered the values (Kumar and E. Futschik, 2007). The centered normalized values were visualized as heat maps, divided into two panels, one for each bacterial strain. Strikingly, heat maps showed obvious clusters of co-expressed genes and clear gene expression differences between adhering (Δ*cps2E*) and non-adhering (wt) bacteria to epithelial cells (**Fig.2CD**). Specifically, the left panel of **Fig.2C** shows the epithelial transcriptional response exposed to encapsulated *S. pneumoniae* at different time points (30, 60, 120 and 240 mpi) while the right panel displays the response in contact with the unencapsulated strain. *Vice versa*, co-expressed clusters in pneumococcal genes are differentially expressed in contact with human epithelial cells (**Fig.2D**).

Making raw data publicly-available has been common practice in recent years, as we have done for this project (GEO accession number GSE79595). Unfortunately, publicly available datasets do not translate to direct exploration and extraction of biological insights for the majority of researchers. Therefore, we built an easily-accessible online platform which hosts the complete dual RNA-Seq database. To access and visualize the data, users can simply select the gene of interest (or multiple genes of interests) and examine their expression during early infection (**Supplementary Figure 3**). Expression data can be downloaded and opened in common spreadsheet software e.g.: Microsoft Excel®. To

visualize expression, users can choose from three normalization methods: DESeq2 normalization (Love et al., 2014), TPM (transcript per million, Wagner et al., 2012) and $\log_2$ -transformed TPM values.

Validation of dual RNA-Seq by qRT-PCR and pneumococcal protein fusions

To confirm dual RNA-Seq data by quantitative real-time PCR (qRT-PCR), we chose 10 epithelial genes (*ABCC2, AKR1B10, AKR1C3, ALDH1A1, DEFB1, DKK1, IDH1, NOLC1, PTGES* and *TXNRD1*) and 19 pneumococcal genes (*amiC, blpY, dinF, hrcA, infC, lytA, malC, manL, msmR, nrdD, pulA,* SPD_0249, SPD_0392, SPD_0475, SPD_0961, SPD_0990, SPD_1517, SPD_1711 and SPD_1798). The target genes were selected because of their varied expression profiles: increasing, decreasing or unchanged. The cycle threshold (Ct) for epithelial transcripts were normalized against *ACTB* (β-actin) while pneumococcal transcripts were normalized to *gyrA* (gyrase A). The reference genes were highly expressed and did not show significant changes (p>0.05, *FC*<2) between any time points during early infection. The qRT-PCR fold change was calculated by the ΔΔCt method (Livak and Schmittgen, 2001) to one time point. Fold changes obtained by qRT-pCR and dual RNA-Seq showed a relatively high correlation for both species: epithelial transcripts, (R^2=0.72) and pneumococcal transcripts, (R^2=0.73, **Fig.3A**), validating the reliability of dual RNA-Seq data.

Since transcript levels do not necessarily correspond with protein expression (Ning et al., 2012; Taniguchi et al., 2010), we quantified four pneumococcal protein levels whose genes showed upregulation during adherence to epithelial cells. We fused a fast-folding variant of the green fluorescent protein (GFP) to the carboxy-termini of SPD_0475, SPD_0963, SPD_1711 and SPD_1716, at their own locus while preserving all upstream

regulatory elements (**Fig.3B**). We transformed these constructs into the *hlpA_hlpA-rfp,*

Δcps2E genetic background (Kjos et al., 2015). SPD_0475 encodes a 204 amino acids (aa)

CAAX amino terminal protease with unknown function. SPD_0963 encodes a 45 aa

hypothetical protein. SPD_1711 (132 aa) was described as a single stranded DNA binding

protein and may assist in competence (Attaiech et al., 2011) and SPD_1716 is a 63 aa

ortholog of cell wall or choline binding protein in other *Streptococcaceae.*

We imaged adherent *S. pneumoniae* with fluorescence microscopy during the

indicated time points (**Fig.3C**). Since (i) RFP serves as an accurate proxy for cell number

and viability (Kjos et al., 2015), (ii) *hlpA* does not change during early infection (p>0.05,

FC<2) and (iii) ratio between the GFP and RFP indicates relative expression of the protein

of interests (red circle and line, **Fig.3D**), we were able to quantify the proteins of interest.

Gene expression values from the dual RNA-Seq data (blue circles and line, **Fig.3D**) show a

degree of correlation with protein level in three out of the four cases – suggesting that

pneumococcal transcriptional changes reflects, to some extent, changes in protein level

(Vogel and Marcotte, 2012).

Pneumococcal ROS induces expression of glutathione-mediated detoxification genes in epithelial cells

Along with pneumococcal adherence and multiplication, we aimed to recapitulate the host

response in our model. We hypothesized that the epithelial transcriptome adapts in

response to bacterial presence, independent of adherence. To identify the responsive

genes, we automatically clustered epithelial working libraries exposed to wild type

pneumococci. 242 epithelial genes were co-expressed in a similar manner (**Fig.4A**), i.e.,

lowly expressed at 30 mpi then sustained upregulation thereafter. Gene ontology (GO)

enrichment (Dennis et al., 2003) indicated that 26 of the subset are associated with oxidation and reduction ($p=5.7 \cdot 10^{-4}$). Moreover, 9 genes are associated directly with glutathione, an ubiquitous antioxidant: *GCLC* and *GCLM*, in glutathione biosynthesis; *GPX2* and *MGST2* in detoxification; and *GSR*, *IDH1*, *IDH2*, *PGD* and *G6PD* in glutathione recycling.

Glutathione, a tripeptide of glutamic acid, cysteine and glycine, is produced and secreted by epithelial cells (Valko et al., 2007). The vital molecule is biosynthesized through amino-acid polymerization and when subjected to ROS (hydrogen peroxide, lipid superoxide or oxygen radicals), glutathione readily donates an electron or hydrogen atom to quench the ROS. The process is assisted by ligands and glutathione peroxidase (GPX2). Oxidized glutathione can be recycled by glutathione reductase (GSR) dependent on NADPH (**Fig.4B**). Alternatively, glutathione conjugates and neutralize ROS (Forman et al., 2009). Expression of nine glutathione-associated genes showed a sustained significant increase in epithelial cells exposed to encapsulated strain ($p<0.05$, 60 vs. 30 mpi, **Fig.4C**).

To validate the abovementioned finding, we repeated the experiment, isolated total RNA and performed qRT-PCR on four genes: *GPX2*, involved in detoxification and *GSR*, *IDH1* and *PGD*, in glutathione recycling. As expected, we observed significant upregulation of these genes between 30 and 60 mpi (**Fig.4D**). Interestingly, Rai et al (2015) showed that pneumococcal supernatant is sufficient to instigate oxidative damage in A549. Indeed, when epithelial cells were incubated with filtered pneumococcal supernatants, the genes were activated (**Fig.4D**). To establish that pneumococci-derived ROS was behind the response, we added the antioxidant, resveratrol (100 µM) to the epithelial-pneumococcal model and did not observe activation (**Fig.4D**) as shown by (Zahlten et al., 2015).

Adherent *S. pneumoniae* repress epithelial innate immune response

Contrasting epithelial gene expression in response to encapsulated and unencapsulated pneumococci allowed us to specifically identify adherence-responsive genes. In doing so, we identified 271 adherence-responsive (*p*<0.05) epithelial genes, of which 25 genes are activated (fold change, *FC*>2) and 248 repressed (*FC*>2) during early-infection (**Fig.5A**). Two genes, *PTGS2* and *HIST1H4B* showed repression and activation at more than one time point. Subsequently, GO-term enrichment in the subset of repressed genes at 60 mpi showed that "immune response" to be enriched in 16 genes (*p*=$2.3 \cdot 10^{-4}$, **Fig.5B**).

 CXCL8 (*IL8*), encoding interleukin-8, was one of the repressed immunity gene. CXCL8 is a potent chemoattractant for neutrophil and other granulocytes. Interestingly, at 60 mpi, Δ*cps2E*-exposed epithelial cells, expressed 2.5±1.3 less *CXCL8* than epithelial cells exposed to wild type *S. pneumoniae* (**Fig.5C**). This difference was validated by qRT-PCR (**Fig.5D**). Further, we asked whether *CXCL8* repression is an active process or merely mediated by physical adherence. To assess this, we co-incubated heat-inactivated Δ*cps2E* and heat-inactivated wild type with epithelial cells. Note that heat inactivation preserves pneumococcal epitope and protein structures (Hvalbye et al., 1999). *CXCL8* was still significantly repressed by dead Δ*cps2E* but not by dead wild type pneumococci (**Fig.5D**) – suggesting that *CXCL8* repression is independent of viability but dependent to presence of the capsule or to the accessibility of surface-exposed (protein) factors in capsule absence. Intriguingly, Graham and Paton (2006) showed that epithelial interleukin-8 production and release was suppressed by pneumococcal surface protein CbpA and incubation with Δ*cbpA* leads to higher CXCL8 expression. We speculate that the absence of capsule in Δ*cps2E* increases accessibility of pneumococcal surface-exposed factors, including CbpA to epithelial receptors, leading to repression of *CXCL8*.

DEFB1, encoding β-defensin-1 an important epithelial-derived constitutively-

expressed antimicrobial peptide (Krisanaprakornkit et al., 1998), is another repressed

immunity gene. However, in our model, *DEFB1* was repressed 3.0±1.2 times ($p<0.05$, 60

mpi) in Δ*cps2E*-exposed epithelial cells compared to wild type-exposed cells (**Fig.5C,**

validated in **Fig.5D**). Additionally, while heat-inactivated wild type pneumococci stimulate

comparable levels of *DEFB1* compared to viable wild type bacteria, non-viable Δ*cps2E*

repressed *DEFB1* expression even lower ($p<0.05$) than viable Δ*cps2E* (**Fig. 5D**). We

conclude that *DEFB1* expression is affected by adherence, accessibility of pneumococcal

surface proteins and pneumococcal viability by an as of yet unknown mechanism. In

summary, we show that adherent pneumococci modulate epithelial expression of innate

immunity genes, including *CXCL8* and *DEFB1*, mediated by pneumococcal surface factors.

Adherent *S. pneumoniae* activate sugar-importers

Consistent to the epithelial analysis, we contrasted pneumococcal gene expression

between unencapsulated and encapsulated libraries, resulting in 248 differentially

expressed ($p<0.05$) genes. Specifically, 110 pneumococcal genes of the 248 were

activated ($FC>2$) in the unencapsulated strain while 133 genes were repressed ($FC>2$,

Fig.6A). Five genes, SPD_0188, SPD_0226, SPD_0415, SPD_1717 and SPD_1988,

showed repression and activation at different time points. We used gene classes developed

for *S. pneumoniae* TIGR4 (van Opijnen and Camilli, 2012) to categorize adherence-

responsive genes (**Fig.6B**). Excitingly, most of the adherence-responsive genes are of

unknown function (133 of 248 genes, 54%), highlighting our paucity of knowledge in

pneumococcal gene function. A large part of the subset (36 genes, 15%) are involved in

cellular transport (**Fig.6B**). Since the pneumococcal genome has an exceptionally high

number of carbohydrate transporters (Bidossi et al., 2012), it is not surprising that 15 of the genes encode for sugar transporters (**Fig.6C**).

At 60 mpi, 11 carbohydrate transporters were differentially expressed ($p<0.05$, $FC>2$) between $\Delta cps2E$ and encapsulated pneumococci exposed to epithelial cells. In the presence of high glucose ($2g \cdot L^{-1}$), $\Delta cps2E$ expressed 1.5 fold less *manLM,* encoding glucose transporters, than encapsulated *S. pneumoniae* (**Fig.6D**)*.* Moreover, seven importers were activated in the adherent strain compared to the free-floating encapsulated strain. The seven non-glucose transporter-genes and their substrates are SPD_0089 (disaccharides: galactose, mannose, N-acetylmannosamine), *celC* (disaccharides: cellobiose, gentiobiose), SPD_0232/33/34 (disaccharides: cellobiose), *rafE* (oligosaccharides: raffinose, stachyose, melliobiose) and *malD* (polysaccharides: maltotitriol, maltodextrine, glycogen, **Fig.6D**). At the same time, four genes were repressed in unencapsulated *S. pneumoniae* exposed to human epithelial cells, *glpF* (glycerol) and SPD_0740/41/42 (ribonucleosides). We selected three transporter-genes, *malD* (polysaccharides), *rafE* (oligosaccharides) and SPD_0234 (disaccharides) and validated the abovementioned observations by qRT-PCR (**Fig.6E**).

Our data indicate that adherent unencapsulated bacteria detect non-glucose sugars in their immediate vicinity. Epithelial mucus may provide non-glucose carbohydrates (Yesilkaya et al., 2008) and simultaneously limit the interaction of epithelial cells to encapsulated wild type bacteria (Nelson et al., 2007). We, then, removed epithelial-associated mucus by washing the surface with warm PBS and observed that the genes were no longer activated ($FC<2$, **Fig.6E**). Next, we incubated pneumococcal strains with type III porcine mucin ($5 \ g \cdot L^{-1}$), mimicking complex carbohydrate in the medium. Interestingly, the importers were not differentially expressed between strains ($FC<2$),

330 indicating similar access to non-glucose sugars (**Fig.6E**). We conclude that following

331 adherence, *S. pneumoniae* senses an enriched host-derived non-glucose carbohydrate and

332 in turn, activates transporters to import the now-available sugars.

Discussion

Early infection is a chaotic and disruptive encounter between host and pathogen. In both

species, a multitude of transcriptional-mediated cellular processes are fine-tuned: activated,

maintained and repressed to ensure survival. The recently-described dual RNA-Seq

approach allows simultaneous host-pathogen monitoring during their interactions (Baddal et

al., 2015; Dillon et al., 2015; Westermann et al., 2016). In this study, we exploited the dual

RNA-Seq approach by application to a pneumococcal infection model to human lung

alveolar epithelial cells. We have generated a detailed time-resolved dataset of epithelial-

pneumococcal transcriptomes up to 4 hours after infection. Moreover, we have validated

the rich dataset by qRT-PCR and quantitative fluorescence microscopy. Furthermore, we

have shown that adherence-specific transcriptional responses in host and pathogen can be

identified by contrasting the transcriptomes of the encapsulated libraries with its

unencapsulated counterpart.

Our early-infection model recapitulated three major *in vivo* characteristics of

pneumococcal infection: pneumococcal adhesion, bacterial multiplication and epithelial

responses to pathogenic burden. Our model recapitulated these infection characteristics: (i)

adherence for both encapsulated and unencapsulated pneumococci (**Fig.1**), (ii)

pneumococcal viability and growth during early infection, e.g.: generation of ROS,

expression of carbohydrate importers (**Figs.4** and **6**) and (iii) host response to

pneumococcal presence, e.g.: glutathione-associated detoxification and innate immune

response (**Figs.4** and **5**). Further, as shown by regulation of carbohydrate transporters (**Fig.6**), pneumococci sensed epithelial presence and subsequently adapted their transcriptome.

Remarkably, we observed almost all (99%) pneumococcal genes are expressed at early infection (**Fig.2**) – in line with recent studies on bacterial transcriptomes adapting to multiple conditions (Kröger et al., 2013; Nicolas et al., 2012). We speculate that interspecies interaction necessitates massive pneumococcal transcriptional adaptation. Moreover, we have observed activation of detoxification genes (*GPX2* and *GSR*) in epithelial cells protecting against pneumococci-derived ROS (**Fig.4**). Indeed, *S. pneumoniae* has been reported to secrete high levels of peroxides as a by-product of its pyruvate metabolism (Carvalho et al., 2011) and has recently been shown to cause DNA-damage-dependent apoptosis in alveolar lung epithelial cells (Rai et al., 2015). In addition, pneumococcal early competence genes (Martin et al., 2000) were activated in our model (http://dualrnaseq.molgenrug.nl competence subset). Future work should also examine whether small non-coding RNAs play a role in pneumococcal early infection as they do in Salmonella (Westermann et al., 2016), something that was not explored currently due to the poor D39 genome annotation in this regard.

Our approach can be expanded further by incorporating relevant LRTI agents into the model. For example, alveolar macrophages and epithelial cells, together, form epithelium lining of the lower respiratory tract. The cells reciprocally influence cellular phenotypes and behaviors (Hussell and Bell, 2014), highly relevant to infection. Moreover, pneumococcal co-infection and secondary infection are not unheard of. Influenzae virus potentiates *S. pneumoniae* by neuraminidase activity that exposes cryptic epithelial receptor and facilitates pneumococcal adherence (Siegel et al., 2014). In addition,

influenzae infection causes loss of superficial epithelial cells, revealing basal epithelium (Kash et al., 2011) and aggravating secondary pneumococcal infection. On the other hand, *H. influenzae* negatively affects pneumococci by recruiting neutrophils and stimulating the killing of opsonized pneumococci (Lysenko et al., 2010). Incorporation of other agents into the model and exploiting dual (or triple, quadruple) RNA-Seq approaches may provide novel insights into respiratory infection.

However, complementary approaches, are needed to obtain a full picture of early infection. Though transcriptome rewiring is a focal point during interspecies interaction (Jenner and Young, 2005; Sorek and Cossart, 2010), non-transcriptional regulation plays an important part during early infection. Capsule shedding, a hallmark of pneumococcal infection, is regulated by autolysin-A (Lyt-A). LytA is activated when the bacterium encounters alveolar cathelicidin (Kietzman et al., 2016), which is independent of transcriptional regulation. Heterogeneity of cellular responses is another confounding factor (Jørgensen et al., 2013). Recently, dual RNA-Seq combined with cell sorting was used to identify heterogeneous activity of Salmonella virulence factor that, in turn, drives heterogeneous interferons response in macrophage (Avraham et al., 2015). This highlights the relevance of noise in gene expression and cell-to-cell variability in host-pathogen interaction. Furthermore, whole organism infection models offers a more systemic perspective. Dual RNA-Seq approach has been used to monitor infection in wheat-bacteria (Camilios-Neto et al., 2014) and mosquito-filaria (Choi et al., 2014). Whole organism dual RNA-Seq is not without its challenges, including averaging (host) effects to gene expression across all cell types.

Besides its relevance in communicable diseases, gained insights into pneumococcal infection are also applicable to understanding several non-communicable respiratory

diseases. Asthma, the most common chronic respiratory disease is a major risk factor for pneumococcal infection (Talbot et al., 2005). Additionally, COPD (Chronic Obstructive Pulmonary Disease, Decramer et al., 2012) and cigarette smoking (Phipps et al., 2010) have been reported to increase the risk of pneumococcal LRTI. Here, we reported the first study to show simultaneous transcriptomic changes of the pathogen *S. pneumoniae* and human lung alveolar epithelial cells during early infection. Additionally, we have made the time-resolved dual transcriptomics dataset available to the broader research community (http://dualrnaseq.molgenrug.nl). We invite pneumococcal researchers to use the database to formulate research questions in the development of preventive and curative strategies against pneumococcal infection. To conclude, we call researchers from the fields of microbiology, immunology and pulmonology to access the dataset and use it to develop their own hypotheses.

Methods

Culture of epithelial cell line, *S. pneumoniae* D39 and pneumococcal transformation

Human type II lung epithelial cell line, A549 (ATCC® CCL-185) and *S. pneumoniae* D39 were routinely cultured without antibiotics. Strain construction is described in detail in **Supplementary information**. Oligonucleotides are listed in **Supplementary Table 1** and strains in **Supplementary Table 2**.

Infection studies

Confluent monolayer of A549 was co-incubated with *S. pneumoniae* D39 at multiplicities of infection 10, in 1% fetal bovine serum in RPMI1640 without phenol red. Prior to infection, epithelial monolayer was kept for 10 more days after confluence. To optimize cell-to-cell contact, centrifugation was employed (2000 ×g, 5 min, 4°C). Adherence assays were

performed by enumeration of plated cfu in blood agar. More in **Supplementary information**.

Simultaneous total host-pathogen RNA isolation

Before RNA isolation, samples were treated by saturated ammonium sulfate solution (Korfhage et al., 2002). Cells were disrupted by bead beating, and total RNA was isolated as described in the **Supplementary information**.

Library preparation, sequencing, data pipeline and online database

Total RNA was dual rRNA-depleted, reverse-transcribed and sequenced on the Illumina NextSeq 500 in 75 nt single end mode. Raw reads were trimmed, aligned into chimeric human-pneumococcus genome. Reads were counted and differential gene expression analysis were performed to epithelial and pneumococcal libraries separately. Automatic clustering and GO enrichment were performed. The online database is built using in-house script. Significantly differentially expressed epithelial and pneumococcal genes are listed in **Supplementary Table 3** and **4**, respectively.

qRT-PCR and quantitative fluorescence microscopy

Infection studies were repeated, total RNA isolated and qRT-PCR is performed. For fluorescence microscopy, infection studies were performed in 8-wells µ-slides (Ibidi, DE). More in **Supplementary information**.

Accession Numbers

The raw data is accessible at http://www.ncbi.nlm.nih.gov/geo/ with accession number GSE79595.

Author Contribution

R.A and J.W.V. designed the research, analyzed the data and wrote article. R.A performed research, J.S analyzed the data, S.H. built online database.

Acknowledgements

We thank W.J. Quax and R. Setroikromo (UMCG, Groningen) for the human cell line, V. Benes and B. Haase (GeneCore, EMBL, Heidelberg) for sequencing support, M. Kjos for fruitful discussions and A. de Jong for bioinformatics support. We thank S. El Aidy and L.E. Keller for comments on the manuscript. Work in Veening lab is supported by the EMBO Young Investigator Program, VIDI fellowship (864.12.001) from the Netherlands Organization for Scientific Research, Earth and Life Sciences (NWO-ALW), and ERC Starting Grant 337399-PneumoCell.

Conflict of Interest

None

References

Abeyta, M., Hardy, G.G., and Yother, J. (2003). Genetic alteration of capsule type but not PspA type affects accessibility of surface-bound complement and surface antigens of *Streptococcus pneumoniae*. Infect. Immun. *71*, 218–225.

Attaiech, L., Olivier, A., Mortier-Barrière, I., Soulet, A.-L., Granadel, C., Martin, B., Polard, P., and Claverys, J.-P. (2011). Role of the Single-Stranded DNA–Binding Protein SsbB in Pneumococcal Transformation: Maintenance of a Reservoir for Genetic Plasticity. PLOS Genet *7*, e1002156.

Avraham, R., Haseley, N., Brown, D., Penaranda, C., Jijon, H.B., Trombetta, J.J., Satija, R., Shalek, A.K., Xavier, R.J., Regev, A., et al. (2015). Pathogen Cell-to-Cell Variability Drives Heterogeneity in Host Immune Responses. Cell *162*, 1309–1321.

Baddal, B., Muzzi, A., Censini, S., Calogero, R.A., Torricelli, G., Guidotti, S., Taddei, A.R., Covacci, A., Pizza, M., Rappuoli, R., et al. (2015). Dual RNA-seq of Nontypeable *Haemophilus influenzae* and Host Cell Transcriptomes Reveals Novel Insights into Host-Pathogen Cross Talk. mBio *6*, e01765–15.

Beiter, K., Wartha, F., Hurwitz, R., Normark, S., Zychlinsky, A., and Henriques-Normark, B. (2008). The Capsule Sensitizes *Streptococcus pneumoniae* to α-Defensins Human Neutrophil Proteins 1 to 3. Infect. Immun. *76*, 3710–3716.

Bidossi, A., Mulas, L., Decorosi, F., Colomba, L., Ricci, S., Pozzi, G., Deutscher, J., Viti, C., and Oggioni, M.R. (2012). A Functional Genomics Approach to Establish the Complement of Carbohydrate Transporters in *Streptococcus pneumoniae* . PLoS ONE *7*, e33320.

Bogaert, D., De Groot, R., and Hermans, P.W.M. (2004). *Streptococcus pneumoniae* colonisation: the key to pneumococcal disease. Lancet Infect. Dis. *4*, 144–154.

Bolger, A.M., Lohse, M., and Usadel, B. (2014). Trimmomatic: A flexible trimmer for Illumina Sequence Data. Bioinformatics btu170.

Bootsma, H.J., Egmont-Petersen, M., and Hermans, P.W.M. (2007). Analysis of the In Vitro Transcriptional Response of Human Pharyngeal Epithelial Cells to Adherent *Streptococcus pneumoniae*: Evidence for a Distinct Response to Encapsulated Strains. Infect. Immun. *75*, 5489–5499.

Camilios-Neto, D., Bonato, P., Wassem, R., Tadra-Sfeir, M.Z., Brusamarello-Santos, L.C., Valdameri, G., Donatti, L., Faoro, H., Weiss, V.A., Chubatsu, L.S., et al. (2014). Dual RNA-seq transcriptional analysis of wheat roots colonized by *Azospirillum brasilense* reveals up-regulation of nutrient acquisition and cell cycle genes. BMC Genomics *15*, 378.

Carvalho, S.M., Kloosterman, T.G., Kuipers, O.P., and Neves, A.R. (2011). CcpA Ensures Optimal Metabolic Fitness of *Streptococcus pneumoniae*. PLoS ONE *6*, e26707.

Choi, Y.-J., Aliota, M.T., Mayhew, G.F., Erickson, S.M., and Christensen, B.M. (2014). Dual RNA-seq of Parasite and Host Reveals Gene Expression Dynamics during Filarial Worm–Mosquito Interactions. PLOS Negl Trop Dis *8*, e2905.

Decramer, M., Janssens, W., and Miravitlles, M. (2012). Chronic obstructive pulmonary disease. The Lancet *379*, 1341–1351.

Dennis, G., Sherman, B.T., Hosack, D.A., Yang, J., Gao, W., Lane, H.C., and Lempicki, R.A. (2003). DAVID: Database for Annotation, Visualization, and Integrated Discovery. Genome Biol. *4*, P3.

Dillon, L.A.L., Suresh, R., Okrah, K., Corrada Bravo, H., Mosser, D.M., and El-Sayed, N.M. (2015). Simultaneous transcriptional profiling of *Leishmania major* and its murine macrophage host cell reveals insights into host-pathogen interactions. BMC Genomics *16*, 1108.

Dobin, A., Davis, C.A., Schlesinger, F., Drenkow, J., Zaleski, C., Jha, S., Batut, P., Chaisson, M., and Gingeras, T.R. (2013). STAR: ultrafast universal RNA-seq aligner. Bioinforma. Oxf. Engl. *29*, 15–21.

Forman, H.J., Zhang, H., and Rinna, A. (2009). Glutathione: Overview of its protective roles, measurement, and biosynthesis. Mol. Aspects Med. *30*, 1–12.

Graham, R.M.A., and Paton, J.C. (2006). Differential Role of CbpA and PspA in Modulation of In Vitro CXC Chemokine Responses of Respiratory Epithelial Cells to Infection with *Streptococcus pneumoniae*. Infect. Immun. *74*, 6739–6749.

Gray, R.D., Duncan, A., Noble, D., Imrie, M., O'Reilly, D.S.J., Innes, J.A., Porteous, D.J., Greening, A.P., and Boyd, A.C. (2010). Sputum trace metals are biomarkers of inflammatory and suppurative lung disease. Chest *137*, 635–641.

Hackett, N.R., Butler, M.W., Shaykhiev, R., Salit, J., Omberg, L., Rodriguez-Flores, J.L., Mezey, J.G., Strulovici-Barel, Y., Wang, G., Didon, L., et al. (2012). RNA-Seq quantification of the human small airway epithelium transcriptome. BMC Genomics *13*, 82.

Hallstrand, T.S., Hackett, T.L., Altemeier, W.A., Matute-Bello, G., Hansbro, P.M., and Knight, D.A. (2014). Airway epithelial regulation of pulmonary immune homeostasis and inflammation. Clin. Immunol. Orlando Fla *151*, 1–15.

Hammerschmidt, S., Bergmann, S., Paterson, G.K., and Mitchell, T.J. (2007). Pathogenesis of *Streptococcus pneumoniae* infections :adaptive immunity, innate immunity, cell biology, virulence factors. In Community-Acquired Pneumonia, (Birkhäuser Basel), pp. 139–181.

Hussell, T., and Bell, T.J. (2014). Alveolar macrophages: plasticity in a tissue-specific context. Nat. Rev. Immunol. *14*, 81–93.

Hvalbye, B.K.R., Aaberge, I.S., Løvik, M., and Haneberg, B. (1999). Intranasal Immunization with Heat-Inactivated *Streptococcus pneumoniae* Protects Mice against Systemic Pneumococcal Infection. Infect. Immun. *67*, 4320–4325.

Hyams, C., Camberlein, E., Cohen, J.M., Bax, K., and Brown, J.S. (2010). The *Streptococcus pneumoniae* Capsule Inhibits Complement Activity and Neutrophil Phagocytosis by Multiple Mechanisms. Infect. Immun. *78*, 704–715.

Jenner, R.G., and Young, R.A. (2005). Insights into host responses against pathogens from transcriptional profiling. Nat. Rev. Microbiol. *3*, 281–294.

Jørgensen, M.G., van Raaphorst, R., and Veening, J.-W. (2013). Chapter 6 - Noise and Stochasticity in Gene Expression: A Pathogenic Fate Determinant. In Methods in Microbiology, C.H. and A. Wipat, ed. (Academic Press), pp. 157–175.

Kadioglu, A., Weiser, J.N., Paton, J.C., and Andrew, P.W. (2008). The role of *Streptococcus pneumoniae* virulence factors in host respiratory colonization and disease. Nat. Rev. Microbiol. *6*, 288–301.

Kash, J.C., Walters, K.-A., Davis, A.S., Sandouk, A., Schwartzman, L.M., Jagger, B.W., Chertow, D.S., Li, Q., Kuestner, R.E., Ozinsky, A., et al. (2011). Lethal synergism of 2009

pandurum H1N1 influenza virus and *Streptococcus pneumoniae* coinfection is associated with loss of murine lung repair responses. mBio *2*.

Kietzman, C.C., Gao, G., Mann, B., Myers, L., and Tuomanen, E.I. (2016). Dynamic capsule restructuring by the main pneumococcal autolysin LytA in response to the epithelium. Nat. Commun. *7*, 10859.

Kjos, M., Aprianto, R., Fernandes, V.E., Andrew, P.W., van Strijp, J.A.G., Nijland, R., and Veening, J.-W. (2015). Bright fluorescent *Streptococcus pneumoniae* for live-cell imaging of host-pathogen interactions. J. Bacteriol. *197*, 807–818.

Korfhage, C., Wyrich, R., and Oelmuller, U. (2002). Ammonium sulfate for neutralization of inhibitory effects (Google Patents).

Krisanaprakornkit, S., Weinberg, A., Perez, C.N., and Dale, B.A. (1998). Expression of the peptide antibiotic human beta-defensin 1 in cultured gingival epithelial cells and gingival tissue. Infect. Immun. *66*, 4222–4228.

Kröger, C., Colgan, A., Srikumar, S., Händler, K., Sivasankaran, S.K., Hammarlöf, D.L., Canals, R., Grissom, J.E., Conway, T., Hokamp, K., et al. (2013). An Infection-Relevant Transcriptomic Compendium for *Salmonella enterica* Serovar Typhimurium. Cell Host Microbe *14*, 683–695.

Kukurba, K.R., and Montgomery, S.B. (2015). RNA Sequencing and Analysis. Cold Spring Harb. Protoc. *2015*, pdb.top084970.

Kumar, L., and E. Futschik, M. (2007). Mfuzz: A software package for soft clustering of microarray data. Bioinformation *2*, 5–7.

Lee, H.-Y., Andalibi, A., Webster, P., Moon, S.-K., Teufert, K., Kang, S.-H., Li, J.-D., Nagura, M., Ganz, T., and Lim, D.J. (2004). Antimicrobial activity of innate immune molecules against *Streptococcus pneumoniae, Moraxella catarrhalis* and nontypeable *Haemophilus influenzae*. BMC Infect. Dis. *4*, 12.

Liao, Y., Smyth, G.K., and Shi, W. (2014). featureCounts: an efficient general purpose program for assigning sequence reads to genomic features. Bioinforma. Oxf. Engl. *30*, 923–930.

Livak, K.J., and Schmittgen, T.D. (2001). Analysis of relative gene expression data using real-time quantitative PCR and the 2(-Delta Delta C(T)) Method. Methods San Diego Calif *25*, 402–408.

Love, M.I., Huber, W., and Anders, S. (2014). Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. Genome Biol. *15*, 550.

Lysenko, E.S., Lijek, R.S., Brown, S.P., and Weiser, J.N. (2010). Within-Host Competition Drives Selection for the Capsule Virulence Determinant of *Streptococcus pneumoniae*. Curr. Biol. *20*, 1222–1226.

Martin, B., Prudhomme, M., Alloing, G., Granadel, C., and Claverys, J.P. (2000). Cross-regulation of competence pheromone production and export in the early control of transformation in *Streptococcus pneumoniae*. Mol. Microbiol. *38*, 867–878.

Mlacha, S.Z.K., Romero-Steiner, S., Hotopp, J.C.D., Kumar, N., Ishmael, N., Riley, D.R., Farooq, U., Creasy, T.H., Tallon, L.J., and Liu, X. (2013). Phenotypic, genomic, and transcriptional characterization of *Streptococcus pneumoniae* interacting with human pharyngeal cells. BMC Genomics *14*, 383.

Nelson, A.L., Roche, A.M., Gould, J.M., Chim, K., Ratner, A.J., and Weiser, J.N. (2007). Capsule enhances pneumococcal colonization by limiting mucus-mediated clearance. Infect. Immun. *75*, 83–90.

Nicolas, P., Mäder, U., Dervyn, E., Rochat, T., Leduc, A., Pigeonneau, N., Bidnenko, E., Marchadier, E., Hoebeke, M., Aymerich, S., et al. (2012). Condition-dependent transcriptome reveals high-level regulatory architecture in *Bacillus subtilis*. Science *335*, 1103–1106.

Ning, K., Fermin, D., and Nesvizhskii, A.I. (2012). Comparative analysis of different label-free mass spectrometry based protein abundance estimates and their correlation with RNA-Seq gene expression data. J. Proteome Res. *11*, 2261–2271.

van Opijnen, T., and Camilli, A. (2012). A fine scale phenotype-genotype virulence map of a bacterial pathogen. Genome Res. *22*, 2541–2551.

Pedersen, M., Nissen, S., Mitarai, N., Svenningsen, S.L., Sneppen, K., and Pedersen, S. (2011). The Functional Half-Life of an mRNA Depends on the Ribosome Spacing in an Early Coding Region. J. Mol. Biol. *407*, 35–44.

Phipps, J.C., Aronoff, D.M., Curtis, J.L., Goel, D., O'Brien, E., and Mancuso, P. (2010). Cigarette Smoke Exposure Impairs Pulmonary Bacterial Clearance and Alveolar Macrophage Complement-Mediated Phagocytosis of *Streptococcus pneumoniae*. Infect. Immun. *78*, 1214–1220.

Prina, E., Ranzani, O.T., and Torres, A. (2015). Community-acquired pneumonia. The Lancet *386*, 1097–1108.

Rai, P., Parrish, M., Tay, I.J.J., Li, N., Ackerman, S., He, F., Kwang, J., Chow, V.T., and Engelward, B.P. (2015). *Streptococcus pneumoniae* secretes hydrogen peroxide leading to DNA damage and apoptosis in lung cells. Proc. Natl. Acad. Sci. *112*, E3421–E3430.

Rajam, G., Anderton, J.M., Carlone, G.M., Sampson, J.S., and Ades, E.W. (2008). Pneumococcal surface adhesin A (PsaA): a review. Crit. Rev. Microbiol. *34*, 131–142.

Rose, M.C., and Voynow, J.A. (2006). Respiratory tract mucin genes and mucin glycoproteins in health and disease. Physiol. Rev. *86*, 245–278.

Shelburne, S.A., Davenport, M.T., Keith, D.B., and Musser, J.M. (2008). The role of complex carbohydrate catabolism in the pathogenesis of invasive streptococci. Trends Microbiol. *16*, 318–325.

Siegel, S.J., Roche, A.M., and Weiser, J.N. (2014). Influenza promotes pneumococcal growth during coinfection by providing host sialylated substrates as a nutrient source. Cell Host Microbe *16*, 55–67.

Sorek, R., and Cossart, P. (2010). Prokaryotic transcriptomics: a new view on regulation, physiology and pathogenicity. Nat. Rev. Genet. *11*, 9–16.

Soumelis, V., Reche, P.A., Kanzler, H., Yuan, W., Edward, G., Homey, B., Gilliet, M., Ho, S., Antonenko, S., Lauerma, A., et al. (2002). Human epithelial cells trigger dendritic cell mediated allergic inflammation by producing TSLP. Nat. Immunol. *3*, 673–680.

St-Pierre, C., Brochu, S., Vanegas, J.R., Dumont-Lagacé, M., Lemieux, S., and Perreault, C. (2013). Transcriptome sequencing of neonatal thymic epithelial cells. Sci. Rep. *3*, 1860.

Talbot, T.R., Hartert, T.V., Mitchel, E., Halasa, N.B., Arbogast, P.G., Poehling, K.A., Schaffner, W., Craig, A.S., and Griffin, M.R. (2005). Asthma as a risk factor for invasive pneumococcal disease. N. Engl. J. Med. *352*, 2082–2090.

Taniguchi, Y., Choi, P.J., Li, G.-W., Chen, H., Babu, M., Hearn, J., Emili, A., and Xie, X.S. (2010). Quantifying *E. coli* proteome and transcriptome with single-molecule sensitivity in single cells. Science *329*, 533–538.

Tecle, T., Tripathi, S., and Hartshorn, K.L. (2010). Review: Defensins and cathelicidins in lung immunity. Innate Immun. *16*, 151–159.

The ENCODE Consortium (2011). Standards, Guidelines and Best Practices for RNA-Seq. https://www.encodeproject.org/

Tseng, H.-J., McEwan, A.G., Paton, J.C., and Jennings, M.P. (2002). Virulence of *Streptococcus pneumoniae*: PsaA mutants are hypersensitive to oxidative stress. Infect. Immun. *70*, 1635–1639.

Valko, M., Leibfritz, D., Moncol, J., Cronin, M.T.D., Mazur, M., and Telser, J. (2007). Free radicals and antioxidants in normal physiological functions and human disease. Int. J. Biochem. Cell Biol. *39*, 44–84.

Vogel, C., and Marcotte, E.M. (2012). Insights into the regulation of protein abundance from proteomic and transcriptomic analyses. Nat. Rev. Genet. *13*, 227–232.

Voynow, J.A., and Rubin, B.K. (2009). Mucins, Mucus, and Sputum. Chest *135*, 505–512.

Wagner, G.P., Kin, K., and Lynch, V.J. (2012). Measurement of mRNA abundance using RNA-seq data: RPKM measure is inconsistent among samples. Theory Biosci. Theor. Den Biowissenschaften *131*, 281–285.

Westermann, A.J., Gorski, S.A., and Vogel, J. (2012). Dual RNA-seq of pathogen and host. Nat. Rev. Microbiol. *10*, 618–630.

Westermann, A.J., Förstner, K.U., Amman, F., Barquist, L., Chao, Y., Schulte, L.N., Müller, L., Reinhardt, R., Stadler, P.F., and Vogel, J. (2016). Dual RNA-seq unveils noncoding RNA functions in host-pathogen interactions. Nature *529*, 496–501.

Yesilkaya, H., Manco, S., Kadioglu, A., Terra, V.S., and Andrew, P.W. (2008). The ability to utilize mucin affects the regulation of virulence gene expression in *Streptococcus pneumoniae*. FEMS Microbiol. Lett. *278*, 231–235.

663 Zahlten, J., Kim, Y.-J., Doehn, J.-M., Pribyl, T., Hocke, A.C., García, P., Hammerschmidt,
664 S., Suttorp, N., Hippenstiel, S., and Hübner, R.-H. (2015). *Streptococcus pneumoniae*-
665 Induced Oxidative Stress in Lung Epithelial Cells Depends on Pneumococcal Autolysis and
666 Is Reversible by Resveratrol. J. Infect. Dis. *211*, 1822–1830.

668 **Figures and Legends**

Figure 1. The early infection model. Confluent monolayer of alveolar epithelial cell line

(A549) was co-incubated with *S. pneumoniae* strain D39. **A**. We chose five infection time points: 0, 30, 60, 120 and 240 minutes post infection (mpi). **B**. Since adherence is a hallmark of infection, we used an unencapsulated strain, created by disrupting *cps2E* vital in capsule biogenesis. **C**. Δ*cps2E* strain adhered more readily to epithelial cells than its encapsulated parental strain. 30 mpi, Δ*cps2E* (orange bar) showed significantly ($p<0.001$) more adherent cells than its parental strain (cyan bar). Data is presented as mean±SEM. **D**. At 240 mpi, both strains multiplied significantly ($p<0.01$) with no significant difference between strains. **E**. Encapsulated strain has more free-floating than adherent cells while Δ*cps2E* has a higher fraction of adherent bacteria. **F**. After quality-check, low-quality reads were trimmed. Reads were aligned to a synthetic chimeric genome. Aligned reads were counted and classified as epithelial or as pneumococcal counts. We removed three gene fractions, clustered and performed functional enrichment to the working libraries.

Fig.2. Dual RNA-Seq generates high-quality datasets suitable for probing host-pathogen transcriptomes. A. On average, there are 70 million reads per library: 18 million reads were mapped to the human genome while 37 million reads to pneumococcal genome. **B.** We excluded three gene fractions: unexpressed genes, i.e.: without counts in any libraries; genes that were differentially expressed at 0 mpi ($p<0.05$) between Δ*cps2E*

and wt libraries and genes with no statistical significance (*p*>0.05) and fold change less than two (*FC*<2) in all contrast. Epithelial working libraries contain 4.009 genes (7% of human genes) while pneumococcal working libraries 868 genes (42% pneumococcal genes). **C.** Gene expression in epithelial working libraries were normalized, centered and automatically clustered. The left panel shows epithelial genes in response to encapsulated strain while the right panel shows epithelial response to Δ*cps2E S. pneumoniae* at different time points. Clear clusters of co-expressed epithelial genes can be observed in the heat map. Blue indicates a relatively lower expression while red, a higher value. **D**. Correspondingly, we presented pneumococcal expression in the same manner: left panel shows wild type pneumococcal response to epithelial cells, while right panel shows Δ*cps2E* response.

Fig.3. Validation of dual RNA-Seq. A. We confirmed dual RNA-Seq gene expression

values by qRT-PCR. The infection study was repeated in duplicates and total RNA was isolated as previously. Ten human and 19 pneumococcal genes were chosen as validation targets. We plotted fold changes from qRT-PCR against dual RNA-Seq fold changes and observed a high degree of correlation ($R^2 > 0.7$, Pearson) for both species. **B.** We also confirmed pneumococcal gene expression at the protein level by quantitative fluorescence microscopy. Four target genes (SPD_0475, SPD_0963, SPD_1711 and SPD_1716) were C-terminally tagged with GFP at their own locus. GFP-fusion were performed in the $\Delta cps2E$ strain expressing RFP fused to HlpA. **C.** Non-deconvolved image of SPD_1711-GFP in $\Delta cps2E$ strain up to 120 mpi. While RFP emitted a relatively constant signal, the GFP signal increases. **D**. We plotted dual RNA-Seq expression values superimposed to the GFP/RFP ratio. To some extent, transcriptional changes corresponded to protein expression.

Fig.4. Epithelial glutathione-associated genes are activated in response to pneumococcal ROS. **A**. We clustered epithelial working libraries exposed to wild type pneumococci and found a cluster of 242 co-expressed genes showing sustained upregulation (*p*<0.05, *FC*>2) at 60 mpi compared to 30 mpi. GO analysis showed that

"oxidation reduction" was enriched in 26 genes (p=5.7·10^{-4}). **B**. Nine genes from the subset are associated with glutathione (GSH), an important antioxidant. Main glutathione-associated processes are biosynthesis of glutathione, detoxification of ROS assisted by ligand and glutathione recycling. **C**. Between 30 and 60 mpi, *GCLC* was increased 2.3±1.1 times and *GCLM* 2.4±1.2 times. *GPX2*, the main detoxification gene was activated 18.7±1.3 times while the ligand, *MGST2* 3.2±1.2 times. Genes involved in the recycling of glutathione were activated: *IDH1,* 4.6±1.2; *IDH2*, 2.4±1.2; *PGD,* 2.9±1.2; *G6PD,* 6.6±1.2 and *GSR,* 2.0±1.1 times. **D**. We validated *GPX2*, *GSR, IDH1* and *PGD* expression with qRT-PCR. Epithelial incubation with pneumococcal supernatant showed similar upregulation of glutathione-associated genes. Addition of resveratrol (100μM) into the model diminished the upregulation (*FC*<2) altogether.

Fig.5. Adherent *S. pneumoniae* repress epithelial innate immune responses. A. At 60 mpi, 97 epithelial genes were significantly repressed upon exposure to Δ*cps2E* bacteria compared to exposure to wild type pneumococci. **B.** GO term enrichment to 60 mpi repressed genes resulted in "immune response" (16 genes, *p*=2.3·10⁻⁴), prostaglandin metabolism (5 genes, *p*=5.6·10⁻⁴) and oxidation reduction (12 genes, *p*=0.019) among others. **C.** Δ*cps2E*–exposed epithelial cells expressed 2.6±1.3 fold <u>less</u> *CXCL8* and 3.0±1.2 fold <u>less</u> *DEFB1* than wild type-exposed epithelial cells. **D.** We validated *CXCL8* and *DEFB1* repression by qRT-PCR. Heat-inactivated encapsulated bacteria showed no repression of *CXCL8* and *DEFB1,* i.e., no difference (*p*>0.05) compared to viable encapsulated *S. pneumoniae* (Dead Encap.). While infection with heat-inactivated Δ*cps2E* repressed *CXCL8* to the level of viable Δ*cps2E*, *DEFB1* was more repressed (*p*<0.05) by

dead Δ*cps2E* than by viable unencapsulated pneumococci.

Fig.6. Adherent pneumococci gain access to host-derived carbohydrates and activate non-glucose sugar importers. A. 248 genes were differentially expressed between pneumococcal strains exposed to epithelial cells: 115 genes of the 248 genes were activated in Δ*cps2E* compared to wt pneumococci while 138 genes were repressed. Note that five genes showed activation and repression at different time points. **B.** Most of the differentially-expressed genes are of unknown function (133 genes, 54% of 248), followed by cellular transport (36 genes, 15%) and DNA replication, repair and recombination (17 genes, 7%). **C.** 15 of the 36 transporter genes are described to transport carbohydrate. The carbohydrate-importers transport a wide range of carbohydrates, from

simple monosaccharides to complex polysaccharides. **D.** At 60 mpi, the expression of glucose transporters (*manLM*, blue boxes) is repressed ($p < 0.05$, *FC*=1.5) in $\triangle cps2E$ compared to encapsulated *S. pneumoniae*. Seven non-glucose transporters are activated ($p < 0.05$, *FC*>2) in the $\triangle cps2E$ strain: SPD_0089, *celC,* SPD_0232/33/34*, rafE* and *malD*. **E.** We validated the data by qRT- PCR for three sugar importers: *malD* (polysaccharides)*, rafE* (oligosaccharide), and SPD_0234 (non-glucose disaccharide). By removing epithelial mucus prior to infection, the importers were no longer activated in $\triangle cps2E$ compared to wild type (*FC*<2, Washed). Incubation with type III porcine mucin ($5\text{g} \cdot \text{L}^{-1}$) did not activate the genes in $\triangle cps2E$ compared to encapsulated pneumococci (*FC*<2).