Multi-layered stochasticity and paracrine signal propagation shape the type-I interferon response

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Introduction

The cellular recognition of viruses evokes the secretion of type-I interferons (IFNs) that induce an antiviral protective state. By live-cell imaging, we show that key steps of virus-induced signal transduction, IFN-β expression, and induction of IFN-stimulated genes (ISGs) are stochastic events in individual cells. The heterogeneity in IFN production is of cellular—and not viral—origin, and temporal unpredictability of IFN-β expression is largely due to cell-intrinsic noise generated both upstream and downstream of the activation of nuclear factor-κB and IFN regulatory factor transcription factors. Subsequent ISG induction occurs as a stochastic all-or-nothing switch, where the responding cells are protected against virus replication. Mathematical modelling and experimental validation show that reliable antiviral protection in the face of multi-layered stochastic cellular heterogeneity is achieved by paracrine response amplification. Achieving coherent responses through intercellular communication is likely to be a more widely used strategy by mammalian cells to cope with pervasive stochasticity in signalling and gene expression.

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The type-I interferon (IFN) system provides a powerful defence against viral infections (Kunzi and Pitha, 2003; Sadler and Williams, 2008; Takeuchi and Akira, 2009). Pathogen recognition receptors, such as RIG-I, stimulate an intracellular signalling cascade that leads to the activation of nuclear factor-κB (NF-κB) and IFN regulatory factors (IRFs) 3 and 7 (Brennan and Bowie, 2010). Subsequent induction of the IFN-β gene and other type-I IFNs is a hallmark of the early response to infection (Theofilopoulos et al, 2005). Upon secretion and binding of type-I IFNs to their specific receptor, the Jak/STAT signalling pathway is activated to reprogram gene expression (Brierley and Fish, 2005). By activation of a wide set of these genes, IFNs act directly by establishing an antiviral state and indirectly through recruiting innate and adaptive immune cells.

On the one hand, IFN production must be tightly regulated to avoid harmful inflammation and autoimmune disease (Trinchieri, 2010). On the other hand, pathogenic viruses inhibit the expression of IFNs or IFN-stimulated genes (ISGs) by diverse mechanisms, indicating that maintaining the efficiency of the IFN system in the face of these viral challenges might be an overriding objective of its evolution (Versteeg and Garcia-Sastre, 2010). Therefore, the finding that only a fraction of cells in a virus-infected cell population expresses IFNs has been surprising (Zawatzky et al, 1985; Hu et al, 2007). Biochemical studies have suggested that stochastic IFN induction results from host cell-intrinsic causes, such as a mechanism of IFN gene induction (Hu et al, 2007; Apostolou and Thanos, 2008) or cellular heterogeneity in expression of the viral sensor RIG-I (Hu et al, 2011). Alternatively, heterocellularity of IFN expression could be caused by the infecting virus (Chen et al, 2010; Killip et al, 2011). It is conceivable that several factors might shape the dynamics of IFN induction, depending on host cell type and virus. Live imaging provides a suitable tool to study dynamics and variability at single-cell resolution (Spiller et al, 2010), but, such an analysis has not yet been carried out for the IFN system.

To understand the functional consequences of cell-to-cell variability in IFN induction, the cellular response to IFNs, particularly the expression of ISGs, must also be defined. Two recent quantitative analyses of IFN-stimulated signalling have modelled the dynamics at the cell-population level (Maiwald
et al., 2010) and considered effects of cell-to-cell variability (Levin et al., 2011). However, the expression of ISGs and the resulting cell fate have not yet been characterized at single-cell level.

Here, we study in living cells and at single-cell resolution both type-I IFN induction and the cellular response to secreted IFN. Using fluorescent reporters based on bacterial artificial chromosomes (BACs) and chimeric transcription factors (Supplementary Figure S1), we imaged successive key steps of IFN induction and response in a prototypical model system, the infection of murine cells in culture with a single-stranded RNA virus. To link our single-cell data to antiviral protection at the cell-population level, we developed a mathematical model based on these data and tested its predictions experimentally.

Our results show that cell-to-cell heterogeneity is a pervasive feature of the IFN system. This heterogeneity manifests itself not only in the virus-induced expression of IFN but also in the IFN-induced protective response. It is to a large part due to cell-intrinsic stochasticity occurring in three key steps: virus-induced signalling, IFN gene induction, and expression of ISGs. In contrast to gene-expression noise described in bacteria and yeast that causes protein levels to fluctuate around a mean value in individual cells (Raj and van Oudenaarden, 2008), the stochasticity found here in the IFN system is an all-or-nothing phenomenon: cells switch on signalling or gene expression or not, and the switching cells do so at widely variable time points. Nevertheless, we find that a reliable antiviral response is achieved through powerful paracrine propagation of the signal. Thus, our results show that the functional dynamics of the IFN response must be understood in terms of the collective spatio-temporal dynamics of stochastically reacting single cells.

Results

Cell-to-cell heterogeneity in IFN induction

In order to monitor authentic IFN-β expression in real time, we stably transfected murine fibroblasts with a BAC-encoded reporter expressing TurboGFP under the control of the Ifnb promoter (IFN-β–tGFP). A representative cell clone was selected showing stable expression of the reporter. These cells were infected with Newcastle Disease Virus (NDV), which replicates and induces IFN in the cells via the double-stranded RNA sensor RIG-I (Kato et al., 2005), without viral interference with this pathway (Childs et al., 2007). As the newly generated viral particles cannot re-infect the mouse cells (Rott, 1979), this system allows us to study in a controlled manner the IFN induction elicited by the primary infection. To quantitatively determine the kinetics and dose response of IFN-β–tGFP expression, reporter cells were infected with NDV and subjected to flow cytometry (Figure 1A). We observed a rise in IFN-β–tGFP-positive cells that faithfully reflected the accumulation of type-I IFNs in the supernatant (Figure 1B). The fraction of IFN producers increased nearly linearly over a broad range of NDV dose, whereas the average expression level already reached ~70% of its maximal value at very low NDV dose (Figure 1C). The frequency of IFN-β expression was very similar for different clones, showing the highly reproducible behaviour of the BAC-based IFN-β–tGFP reporter (Supplementary Figure S2). Thus, the production of IFN-β in response to viral infection is controlled by the fraction of responding cells.

To examine whether IFN expression correlates with viral replication, we jointly measured the viral protein, hemagglutinin-neuraminidase (HN), and IFN-β–tGFP by flow cytometry. In agreement with previous findings (Kumagai et al., 2009; Rehwinkel et al., 2010), only cells with replicating virus expressed IFN-β–tGFP. However, a large proportion of cells with replicating virus did not activate the Ifnb promoter. Strikingly, we observed no correlation between the extent of replication and the fraction of IFN-β–tGFP-expressing cells (Figure 2A and B; Supplementary Figure S3). These observations suggest that the presence of replicating virus in a cell is necessary but not sufficient to induce IFN-β.

As an alternative explanation of heterogeneous IFN expression, it has been suggested that defective viruses are primarily responsible for inducing IFN during parainfluenza virus type 5 (PIV5) infection while during normal replication of PIV5 effective pathogen-associated molecular patterns (PAMPs) are not produced or exposed (Killip et al., 2001). Therefore, we examined the intracellular RNA of non-responding virus-infected cells for its ability to induce IFN in naive cells (Figure 2C; Supplementary Figure S4). RNA from non-responding cells induced IFN to a comparable extent as RNA from responding cells. This finding shows that the viral RNA in IFN-producing and non-producing cells is equally capable of inducing IFN and thus the heterogeneity of IFN-β induction occurs despite the presence of IFN-inducing viral RNA.

Time-lapse microscopic data revealed that the onset of IFN-β expression after infection of cells with NDV varied strongly among the IFN-β producers, with cell-to-cell differences as large as 20 h (Figure 3A; Supplementary Figure S11; Supplementary Movie S1). While the increase in viral load results in earlier onset, the relative temporal variability of IFN-β–tGFP expression differs only slightly. To dissect whether this variability is due to variable time points of infection or reflects cell-intrinsic properties, we bypassed viral infection by liposome-transfecting the cells with the dsRNA analogue poly I:C (liposome-free delivery of poly I:C did not lead to IFN-β expression). Among the IFN producers and independent of the poly I:C concentration, the onset time of IFN-β expression varied strongly and was quantitatively comparable to the case of viral infection, as seen by the same order of magnitude of the coefficient of variation (CV) (Figure 3A). Furthermore, unchanged temporal variability was obtained when synchronizing viral entry by a temperature shift (Supplementary Figure S5). Our data show that the IFN-β gene is induced with widely varying time delays in the producing cells that are not due to variable infection times. In summary, we conclude that the cell-to-cell heterogeneity in IFN-β expression is predominantly of cellular origin.

Both virus-induced signal transduction and Ifnb gene expression are sources of heterogeneity

To analyse mechanistically how cell-to-cell heterogeneity in IFN induction arises, we monitored the activation of the key transcription factors NF-κB and IRF-7 in dual reporter cells,
representing the IKKα/β/γ–NF-kB and IKKα/β–IRF-3/IRF-7 pathways downstream of the dsRNA sensor RIG-I. Fluorescent protein-tagged NF-kB and IRF-7 (p65–YFP and IRF-7–CFP) were localized predominantly in the cytoplasm in uninfected cells and accumulated in the nucleus after NDV infection (Figure 3B). Nuclear translocation of the two transcription factors occurred at the same time in a given cell, but the joint translocation time varied widely between cells (Figure 3C). This finding was corroborated by antibody staining of endogenous IRF-3 and NF-kB (p65), showing that the activation of both factors upon NDV infection or poly I:C stimulation correlated in individual cells (Supplementary Figure S6). Thus, strong cell-to-cell variability arises in the transcription factors occurred at the same time in a given cell, but the joint translocation time varied widely between cells (Figure 3C). This finding was corroborated by antibody staining of endogenous IRF-3 and NF-kB (p65), showing that the activation of both factors upon NDV infection or poly I:C stimulation correlated in individual cells (Supplementary Figure S6). Thus, strong cell-to-cell variability arises in the transcription factors downstream of the dsRNA sensor RIG-I.

To relate IFN induction to the activation of the latent transcription factors we used dual reporter cells expressing the IRF-7–CFP fusion protein and the transcriptional reporter IFN-β–tGFP (Figure 3D; Supplementary Figure S7). We found that the majority of cells translocating IRF-7 to the nucleus also activated the Ifnb promoter (91% at 80 HAU/ml NDV) (Figure 3E). Only few cells (9%) did not express IFN-β–tGFP upon IRF-7 nuclear accumulation, showing the same distribution of IRF-7 translocation times as the IFN-β-expressing cells (Figure 3E, bottom part). No IFN-β–tGFP induction was observed without preceding IRF-7 nuclear accumulation. Thus, IRF-7 nuclear translocation is strictly required for IFN-β expression. The tight coupling of the two events also shows the validity of the reporters used. In agreement with the results in Figure 3C, there was strong temporal heterogeneity in IRF-7 translocation between the cells (signalling delay from viral infection to IRF-7 translocation $T_{\text{sig}} = 11.7 \pm 4.0 \text{ h}$). The interval between IRF-7 translocation and the onset of IFN-β–tGFP gene expression also varied considerably (gene expression delay, $T_{\text{gen}} = 3.4 \pm 1.5 \text{ h}$; note that the CVs are similar, $\sigma_{\text{sig}}/T_{\text{sig}} = 0.34$, $\sigma_{\text{gen}}/T_{\text{gen}} = 0.44$). These quantitative data show that both cytoplasmic signalling from viral entry to the activation of latent transcription factors and induction of IFN-β expression cause strong heterogeneity in IFN-β production.

**Sister-cell analysis reveals cell-intrinsic stochasticity**

Cell-to-cell variability can arise from the intrinsic noise of biochemical reactions and from extrinsic factors, such as...
differences in cell-cycle stage or cellular environment (Elowitz et al., 2002; Maheshri and O’Shea, 2007; Paixão et al., 2007; Raj and van Oudenaarden, 2008; Snijder and Pelkmans, 2011). To minimize extrinsic cell-to-cell differences, we analysed sister cells after division (Spencer et al., 2009). Cells that divided after the 1 h period of infection were followed. IRF-7 activation in sister cells occurred mainly asynchronously, differing by >2 h in ∼50% of cell pairs (Figure 4A). The time between IRF-7–C FP activation and IFN-β–tGFP expression correlated even less in sister cells (Figure 4B), consistent with the previously described stochastic transcription of the Ifnb gene (Apostolou and Thanos, 2008). The coefficient of determination $r^2$ was 0.6 for Tsig and 0.34 for Tgen, indicating that 40% of the variability in signalling and 66% of the variability in Ifnb gene expression are uncorrelated between sister cells and thus provide an estimate for the cell-intrinsic stochasticity. To examine whether different viral replication kinetics in sister-cell pairs are a source of variability, we subjected cells to poly I:C stimulation. Also, under these conditions cells displayed largely uncorrelated IRF-7 signalling and IFN-β gene expression (Figure 4E and F), with similar coefficients of determination ($r^2 = 0.54$ for Tsig and 0.11 for Tgen) as with viral infection. Taken together, these findings show that cell-intrinsic stochasticity is a strong source of cell-to-cell heterogeneity in IFN-β expression.

We checked whether the observed differences between sister cells relate to their time of division. For this purpose, we plotted the time differences between sister cells (ΔTsig and ΔTgen) versus the time elapsed since cell division. The very weak correlations argue against strong control of heterogeneity by the cell cycle (Figure 4C and D for NDV infection and Figure 4G and H for poly I:C stimulation).

Taken together, the sister-cell analysis indicates a role for both ‘extrinsic’ variability between cells and cell-intrinsic stochasticity. The cell-intrinsic component is strong, accounting for approximately half of the variability in the kinetics of antiviral signalling and IFN-β induction in individual cells. This intrinsic stochasticity provides a rationale for the lack of correlation of IFN-β expression with the extent of viral replication (cf. Figure 2A).

**Antiviral protection is an IFN concentration-dependent switch in individual cells**

The intrinsic stochasticity indicates that the responsiveness of IFN-β induction towards virus is not maximized (i.e.,
many cells with replicating virus do not express IFN-β or do so only very late after infection). Therefore, we asked how IFN-β production translates into antiviral protection. We chose IRF-7 as a prototypical ISG (Honda et al., 2005) and measured its expression upon IFN-β stimulation in cells stably transfected with a BAC-encoded IRF-7–mCherry fusion gene.
Flow-cytometric analysis showed a digital pattern of IRF-7–mCherry levels, with distinct expressing and non-expressing subpopulations, where the expressing subpopulation increased with extracellular IFN-β concentration (Figure 5A). The binary IRF-7 response was consistently found for several IRF-7–mCherry clones (Supplementary Figure S8); it was not caused by competition of cells for IFN-β, as IFN-β was still detectable in the supernatant for 4–30 h (Supplementary Figure S9).

To correlate IRF-7–mCherry with the expression of endogenous ISGs, we separated IRF-7–mCherry expressing and non-expressing cells by fluorescence-activated cell sorting (FACS) and subjected isolated RNA from these cells to qPCR analysis. All tested ISGs showed enhanced mRNA levels in IRF-7–mCherry expressing cells (Figure 5B). These findings suggest that a distinct subpopulation of cells that is identified by IRF-7–mCherry expression coordinately expresses an antiviral gene programme. To test this, we stimulated IRF-7–mCherry reporter cells with IFN-β for 8 or 24 h to allow induction of ISGs and then infected the cells with NDV. Virus replicated in IRF-7–mCherry non-expressing cells but not in the expressing cells (Figure 5C). This result shows that, as predicted, only the IRF-7–mCherry positive subpopulation of cells has been protected by IFN. Taken together, these data demonstrate that antiviral protection is a stochastic, IFN-β concentration-dependent, switch in individual cells.

Mathematical modelling predicts paracrine propagation of the IFN response

Our data show a remarkable heterogeneity of type-I IFN expression and subsequent induction of an antiviral state. To understand the resulting dynamics of the IFN system in an infected cell population, we developed a multi-scale mathematical model that describes viral replication and the expression of IFN and ISGs in individual cells together with intercellular communication through secreted IFN (Figure 6A; Supplementary Text S1). According to the sister-cell data, signalling and gene expression contribute strong cell-intrinsic stochasticity (cf. Figure 4). For this reason, we attribute cell-to-cell variability in the model to intrinsic stochasticity in both kinds of processes.

Two sources of heterogeneous IFN induction are the (i) activation of the latent NF-κB and IRF transcription factors downstream of viral sensing by RIG-I and (ii) induction of the Ifnb gene by nuclear NF-κB and IRFs. Importantly, nuclear transcription factor concentration and IFN-β protein increase quantitatively very rapidly once nuclear translocation and gene induction, respectively, have been triggered. By contrast, the waiting times for these events in an individual cell are much longer, so that IFN-β induction is switch-like in time (cf. sample trajectories in Figure 3B and D). Therefore, we describe the dynamics of these events as stochastic transitions $w_R/C_6$ and $w_I/C_6$ between discrete cell states (virus-infected, NF-κB/IRF-activated, IFN-β expressing along the horizontal axis).
Figure 5  Bimodal antiviral response towards IFN. A BAC-based reporter construct containing mCherry fused to the C-terminus of the chromosomal IRF-7 gene was integrated into NIH3T3 cells. Experiments were performed with a cell clone exhibiting a stable integration of the BAC and a representative response towards IFN. (A) Binary dose- and time-dependent IRF-7–mCherry expression. IRF-7–mCherry reporter cells were stimulated with the indicated concentrations of IFN-β, mCherry expression was determined by flow cytometry. (B, C) Bimodality of IRF-7 expression is reflected in ISG transcription and antiviral protection. (B) Reporter cells were treated with 500 U/ml IFN-β for 16 h and subjected to FACS. IRF-7–mCherry positive (2) and negative (1) populations indicated by the shaded areas were separated. RNA was prepared from both populations and analysed by qRT–PCR for expression of the indicated ISGs and Rps9 as a control. The results were normalized to β-actin mRNA and are shown as fold increase of untreated reporter cells. (C) Reporter cells treated with 500 U/ml IFN-β for 8 and 24 h were infected with 80 HAU/ml NDV. In all, 20 h after infection, cells positive (blue) or negative (red) for NDV HN (left column, intracellular antibody staining) were analysed for IFN-stimulated IRF-7–mCherry expression (right column) by flow cytometry.
in Figure 6A) (Mariani et al, 2010). The intracellular replication of virus particles is modelled as a birth–death process. Secreted IFN diffuses in the extracellular space and induces ISGs in an autocrine or paracrine manner. Given the high diffusion coefficient of IFN (Kreuz and Levy, 1965), diffusion on the relevant length scale of several cell diameters takes place within minutes, compared with the hour-timescale of IFN and ISG expression, so that we assume uniform distribution of secreted IFN. To account for the switch-like induction of ISGs, such as IRF-7 (cf. Figure 5), we modelled the IFN-dependent activation rate of STAT1/2 nuclear translocation, \( w_{\text{STAT}} \) (Lillemeier et al, 2001; Vinkemeier, 2004), followed by the rate of ISG induction, \( w_{\text{ISG}} \), resulting in a bimodal distribution of ISG-expressing and non-expressing cells. Virus ceases to replicate in cells expressing ISGs.

To determine the model parameters from the experimental data, we focused on our extensive data for high-dose infections, 40 and 80 HAU/ml NDV (Figure 6; Supplementary Text S1; Supplementary Figure S10). We first computed the distributions for the NF-κB/IRF-7 nuclear translocation times and the IFN-β-gfp switching-on times. Parameter optimization using as objective function the squared sum of residuals between measured and computed (binned) distributions yielded a good fit (Figure 6B and C and green trace in Figure 6D). This fit required that (i) induction of RIG-I-mediated signalling by virus is cooperative (Supplementary Figure S10B) and (ii) Ifnb gene induction is modelled as a multi-step process (Supplementary Figure S10C). Both implications of the quantitative modelling of the data—cooperative RIG-I signalling and multi-step assembly of the IFN-β enhancosome—are consistent with experimental findings (Ford and Thanos, 2010; Onoguchi et al, 2010).

Next, we parameterized IRF-7/ISG induction by fitting the dose response of IRF-7-expressing cells versus external IFN-β (Supplementary Figure S10D) and determining the IFN-β secretion rate per cell to match the observed time course of extracellular IFN-β (Figure 6D, blue trace) to the fraction of IFN-β producers (Figure 6D, green trace). With these parameters (Supplementary Table S1), the simulated kinetics of IRF-7-expressing cells agreed with the data (Figure 6D, red trace; note that IRF-7 induction takes place during prolonged elevation of extracellular IFN-β). Importantly, the stochastic switching in individual cells translates into predictable dynamics of IFN-secreting and protected cell fractions at the population level.

The comparison between model and data in Figure 6D suggests that the IFN-β produced by a single cell can induce ISGs in several cells. To examine this further, we simulated the model for smaller infection doses where IFN-β-producing cells are sparser. Measuring ISG expression, we found good agreement between model and experiment over a range of small doses of infection (Figure 7A). Importantly, the model predicted strong paracrine amplification of antiviral protection through IFN, with up to 40 times as many ISG-expressing cells as IFN-secreting cells. This prediction proved remarkably accurate when tested experimentally (Figure 7B and C). Thus, at the level of the cell population, paracrine amplification transforms stochastic single-cell responses into efficient and predictable antiviral protection. Few IFN-producing cells suffice as sentinels of viral infection to protect a large number of cells surrounding the infection site.

**Discussion**

In this paper, we have dissected the dynamics of the type-I IFN system in an experimental setting that allowed reliable quantitation in living cells. In addition to heterogeneous IFN induction, we describe for the first time bimodal, all-or-nothing expression of ISGs that governs resistance to viral infection. Thus, heterogeneity of single-cell decisions is a pervasive feature of the IFN system. Imaging signal transduction and gene-expression dynamics, we show that these processes contribute quantitatively to heterogeneous expression of the Ifnb gene, and that cell-intrinsic stochasticity prevails in both antiviral signalling and IFN-β expression. Importantly, however, we find that paracrine propagation of the IFN signal can mediate reliable antiviral protection at the cell-population level.

The origin of heterogeneous IFN induction has been controversial; both cell-intrinsic stochastic events in gene expression (Hu et al, 2007; Apostolou and Thanos, 2008) and virus-related mechanisms (Chen et al, 2010; Killip et al, 2011) have been implicated. Our live-cell imaging data obtained with both virus and viral-surrogate (poly I:C) stimuli demonstrate that cell-intrinsic mechanisms generate large-scale heterogeneity in the decision to express IFN-β and in the timing of expression. This establishes intrinsic host cell variability as a decisive factor shaping the antiviral response at the single-cell level while not excluding additional virus effects in other systems (e.g., interference with IFN-β induction, which is a hallmark of many pathogenic viruses).

Cell-intrinsic stochasticity occurs at multiple levels of regulation to generate heterogeneity in IFN-β expression. Quantifying the relative contributions of gene expression and signal transduction activating IRFs and NF-κB, we found that...
both processes have strong stochastic components. The decision whether to express the \textit{Ifnb} gene was made primarily at the level of signal transduction, indicating a limitation in RIG-I-induced signalling. Stochasticity in gene expression contributed to temporal variability. These data show that heterogeneous IFN-\(\beta\) expression cannot be attributed solely to
Stochastic gene regulation, which has previously been the main suspect for IFN-β and other cytokines (Hu-Li et al., 2001; Guo et al., 2004; Apostolou and Thanos, 2008; Raj and van Oudenaarden, 2008; Mariani et al., 2010). Rather, both virus-induced signalling and gene regulation could have different control on cell-to-cell variability in different cell types, depending, for example, on the expression levels of pathway components.

Our model prediction that viral induction of signalling exhibits cooperativity might be linked to the clustering of the adaptor IPS-1 downstream of RIG-I (Onoguchi et al., 2010). Rather, both virus-induced signalling and gene regulation could have different control on cell-to-cell variability in different cell types, depending, for example, on the expression levels of pathway components.

Despite the large heterogeneity of single-cell decisions to express IFN-β and ISGs, the cell-population response is remarkably predictable. Our data-driven model indicates that this is a robust feature of the system that requires simply an appropriate tuning of IFN-β production per cell and IFN responsiveness of ISG induction, allowing paracrine propagation of the IFN signal. In this way, an individual IFN-secreting cell can bring a defined territory surrounding it into an antiviral state.

Thus, heterogeneous and stochastic IFN induction could primarily reflect mechanistic limitations of the host cells (as has been speculated for other cytokines; Mariani et al., 2010) that are readily overcome through paracrine signal propagation. Alternatively, heterogeneity may have functional benefits. In particular, the widely distributed onset times of producers could sustain signalling if the IFN production period per cell is limited. Moreover, we expect that different interference strategies of pathogenic viruses, inhibiting IFN versus ISG induction (Versteeg and Garcia-Sastre, 2010), will have differential effects on the spatio-temporal dynamics of the IFN system which, in turn, might control the viral ‘choice’ between persistent versus acute lytic infection.
Our demonstration of paracrine cell-to-cell communication as an efficient mechanism for quenching single-cell stochasticity will likely be relevant also for other systems. It underscores the importance of studying single cell behaviour within cell-population context for rationalizing cell behaviour.

Materials and methods
Plasmids and BAC constructs
Cloning of BACs was performed by homologous recombination (Zhang et al., 2000) in Escherichia coli strain GS1783 (Tischler et al., 2010). BACs of the RPCI-24 library were obtained from Children’s Hospital Oakland Research Institute. IFN-β-tGFP is based on clone RP24-325J1, where the Ifnb ORF was replaced by TurboGFP ORF (Evrogen) containing an N-terminal triple FLAG epitope and a FRIT-site flanked neomycin phosphotransferase gene under control of the PGK promoter as described earlier (Pulverer et al., 2010). IRF-7-mcherry is based on clone RP24-266F8, in which the IRF7 stop codon and 3′ adjacent ~300 bp were replaced by the mCherry (Shaner et al., 2004) coding sequence and the above-mentioned neomycin selection cassette. Cassettes containing TurboGFP or mCherry together with the selection marker were amplified by PCR introducing 50 bp target sequence homologies (Ifnb–tGFP forward: 5′-GACCTGACGACAGCACGCTTTCCCTCATGACTAAAGACACAGAAGGAG-3′; Ifnb–tGFP reverse: 5′-GATGACACAGACGCTTTCCCTCATGACTAAAGACACAGAAGGAG-3′). Secondary Cy3- or Cy5-labelled antibody (goat anti-mouse IgG, Santa Cruz) was used. Analysis of living and fixed cells in flow cytometry was performed with BD FACsCalibur and LSR II (analytical) or FACsaria (sorting), using appropriate laser and filter settings for TurboGFP, mCherry, and chemical fluorophores. Results were quantified with FlowJo 7.6 software.

Cell culture and DNA transfection
Murine fibroblastoid NIH 3T3 cells were grown under standard tissue culture conditions in DMEM medium (Sigma) supplemented with 10% fetal calf serum, glutamine, penicillin and streptomycin. Transfections of BAC or plasmid DNA were performed with Metafectene (Biontex). Transfections were performed with QIAGEN QuantiTect SYBR green kit in accordance with the manufacturer’s instructions. G418- or puromycin-selected representative clonal cells showing stable expression of the reporter construct and strong signal to background ratio of the fluorescent marker were used. Transfections of BAC or plasmid DNA were performed with Metafectene (Biontex).

IFN stimulation
Infections with Newcastle Disease Virus LaSota (Lohmann Tierzucht, Cuxhaven, Germany) were performed after three times washing with serum-free medium. After 1 h, residual virus was removed by washing three times with serum-containing medium. Low molecular weight poly I:C (InvivoGen) was transacted via Lipofectamine2000 (Invitrogen) according to the manufacturer’s instructions. Total RNA of fractionated cell populations was isolated using RNeasy (QIAGEN) according to the manufacturer’s instructions. G418- or puromycin-selected representative clonal cells showing stable expression of the reporter construct and strong signal to background ratio of the fluorescent marker were used. Transfections of BAC or plasmid DNA were performed with Metafectene (Biontex).

IFN source and detection
Murine IFN-β was recombinantly produced in stably expressing BHK-21 cells and obtained from supernatant as described before (Pulverer et al., 2010). IFN activity was measured with a bioassay based on Mx2-Luc reporter gene expression (Kugel et al., 2011).

Time-lapse microscopy
Time-lapse microscopy used an Olympus IX81 fluorescence microscope, with a PRIOR motorized stage and climate chamber. The Olympus Cell^M autofocus function was used to compensate Z-drift. Cells were seeded into ibidi eight-well μ-slides coated with collagen IV. Image acquisition was performed at intervals between 15 and 30 min.

Image and data analysis
Microscopical picture series were analysed with ImageJ (NIH, Bethesda, MD) and built-in plugins as well as MTrackJ (E. Meijering). Time points of p65 or IRF-7 activation were plotted upon significant increase of nuclear signal that eventually led to maximal accumulation. Time of cell division was counted when sister cells were clearly separated. Figures and statistical analysis were done with MATLAB, Microsoft Excel 2007 and GraphPad Prism 5.

Immunocytochemistry and flow cytometry
NDV-infected cells were stained in suspension following fixation with 4% paraformaldehyde (10 min) and Triton X-100 (0.1%) permeabilization. Primary mouse anti-HN14f (Santa Cruz) was used at 1:500. After washing with PBS supplemented with saponin (0.1%), secondary Cy3- or Cy5-labelled antibody (goat anti-mouse IgG + IgM, Dianova) was used. Analysis of living and fixed cells in flow cytometry was performed with BD FACsCalibur and LSR II (analytical) or FACsaria (sorting), using appropriate laser and filter settings for TurboGFP, mCherry, and chemical fluorophores. Results were quantified with FlowJo 7.6 software.

Real-time PCR
mRNA was isolated with RNeasy (QIAGEN), cDNA hybrid strands were generated with first-strand cDNA synthesis kit (GE Healthcare). Reactions were performed with QIAGEN QuantiTect SYBR green kit in a Roche LightCycler 480 using specific intron-spanning primers (Mx2: forward 5′-TACCACTGAGTTTACGAG-3′, reverse 5′-CATCTTTCCTTCTCAATC-3′; Mx1: forward 5′-TACCACTGAGTTTACGAG-3′, reverse 5′-CATCTTTCCTTCTCAATC-3′; Mx: forward 5′-TACCACTGAGTTTACGAG-3′, reverse 5′-CATCTTTCCTTCTCAATC-3′; Irf7: forward 5′-CATCTTTCCTTCTCAATC-3′, reverse 5′-TACCACTGAGTTTACGAG-3′, reverse 5′-CATCTTTCCTTCTCAATC-3′; Isg20: forward 5′-CATCTTTCCTTCTCAATC-3′, reverse 5′-TACCACTGAGTTTACGAG-3′, reverse 5′-CATCTTTCCTTCTCAATC-3′; Isg15: forward 5′-CATCTTTCCTTCTCAATC-3′, reverse 5′-TACCACTGAGTTTACGAG-3′, reverse 5′-CATCTTTCCTTCTCAATC-3′; Isg15: forward 5′-CATCTTTCCTTCTCAATC-3′, reverse 5′-TACCACTGAGTTTACGAG-3′, reverse 5′-CATCTTTCCTTCTCAATC-3′; Usp18: forward 5′-TACCACTGAGTTTACGAG-3′, reverse 5′-CATCTTTCCTTCTCAATC-3′; Usp18: forward 5′-TACCACTGAGTTTACGAG-3′, reverse 5′-CATCTTTCCTTCTCAATC-3′; Usp18: forward 5′-TACCACTGAGTTTACGAG-3′, reverse 5′-CATCTTTCCTTCTCAATC-3′; Usp18: forward 5′-TACCACTGAGTTTACGAG-3′, reverse 5′-CATCTTTCCTTCTCAATC-3′).
Supplementary information

Supplementary information is available at the Molecular Systems Biology website (www.nature.com/msb).

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Author contributions: MK, HH, UR, and TH conceived the study. UR, MK, JS, GN, MN, and AK carried out the experiments. MR, MF, KK, and TH performed the modelling. UR, MR, MK, HH, MF, and TH analysed the data. UR, MK, HH, MR, and TH wrote the paper.

Conflict of Interest

The authors declare that they have no conflict of interest.

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