Herpesviral Protein Networks and Their Interaction with the Human Proteome
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Herpesviruses are widely spread throughout vertebrates and have large, double-stranded DNA genomes encoding between about 70 and 170 viral proteins, which is only one order of magnitude less than the number of proteins encoded by bacterial genomes. Although studies have revealed a large number of interactions between herpesviral and host proteins, relatively little is known about interactions among herpesviral proteins, particularly for those herpesviruses that replicate poorly in cell culture, e.g., Kaposi sarcoma–associated herpesvirus (KSHV) (1). We thus generated genomewide protein interaction maps for two human pathogens: KSHV, which is a member of the γ-herpesvirus subfamily associated with Kaposi sarcoma and B cell lymphomas (2), and varicella-zoster virus (VZV), which is a member of the α-herpesvirus subfamily that causes chickenpox and shingles (3).

We cloned the open reading frames (ORFs) of both viruses (currently there are 89 ORFs identified in KSHV and 69 ORFs in VZV) by recombinatorial cloning and generated yeast two-hybrid (Y2H) bait and prey arrays. To circumvent the limitation of the Y2H system for transmembrane proteins, we cloned full-length proteins as well as extra- and intracellular domains separately. In KSHV, we tested more than 12,000 viral protein interactions involving both full-length proteins and protein fragments and identified 123 nonredundant interacting protein pairs (fig. S1 and table S1). To date, only a small number of intraviral protein interactions have been reported for KSHV, of which 71% were captured by our screen (table S2). To further confirm the quality of our Y2H results and generate a set of high-confidence interactions, we tested all positive Y2H interactions in parallel by both β-galactosidase and coimmunoprecipitation (CoIP) assays (fig. S2).

About 50% of the protein interactions could be confirmed by CoIP (table S1), and many of the remaining ones have orthologous interactions in other herpesviruses that could be confirmed (table S3). Although our array-based two-hybrid system is internally controlled, some of the interactions not confirmed by CoIP may be nonphysiological and, for example, caused by autoactivation. A comparison between protein interactions and expression profiles of KSHV-infected cells indicated that protein interactions occurred predominantly between proteins expressed at the same time point after infection (fig. S3). In VZV, we detected 173 nonredundant intraviral protein interactions out of ~10,000 tested bait-prey pairs (table S4).

Although cellular protein interaction networks exist for several model organisms, none of the studies on viral protein interactions published so far produced a large enough data set to constitute a protein interaction network. In many cellular protein interaction networks, most nodes have few neighbors, although some have many interaction partners (so-called hubs). The degree distributions of cellular protein networks were reported to follow a power-law decay, and they have been classified as scale-free (4). Like their cellular counterparts, KSHV and VZV had relatively many hubs, a key characteristic of scale-free networks (Fig. 1A; fig. S4). However, in contrast to known cellular protein interaction networks, in which nodes with a single interaction partner are most abundant, the viral networks had relatively few such “peripheral” nodes lying on the “edge” of the network. In KSHV, for example, the degree distribution peaks at nodes with three neighbors. This unusual characteristic at low node degrees is one of the reasons that viral networks appear as single, highly coupled modules and presumably reflects their incompleteness as stand-alone networks. Whereas cellular protein networks have been shown or assumed to be scale-free, the degree distribution of viral networks does not present such a clear-cut picture (fig. S5). If viral networks are approximated by a power-law distribution, they have unusually small power coefficients distinctive from those of known cellular networks and defy current dynamic network evolution models (Fig. 1B and Table 1).

Another important characteristic of complex networks is the so-called small-world property (5). In a small-world network, the average distance between any two nodes is short (short characteristic path length, or the six degrees of separation phenomenon) and local neighborhoods are more densely connected (high clustering coefficient). Both viruses exhibited a short characteristic path length and a short network diameter (the maximum distance between any two nodes), which also suggests their coupling as single modules (Table 1). To assess the viral levels of local clustering, we generated random networks of the same size and degree distribution. The level of local clustering was low in KSHV and VZV, indeed, comparable to the level in equivalent random networks, and thus these viral networks cannot be classified as small-world. In contrast, most cellular protein interaction networks are unambiguously small-world, even after the effect on local clustering due to degree distribution and network size, which is substantial, is filtered out. For example, the clustering coefficient of the Saccharomyces cerevisiae protein interaction network is increased ~30-fold over simulated networks (Table 1). In S. cerevisiae, Maslov and Sneppen demonstrated that hubs tend to avoid each other while preferring low-connectivity nodes (6). As a result, the yeast network has well-separated modules, and errors in one module do not easily propagate to other modules. In the viral networks, there was no such declining degree correlation, and hubs did not tend to avoid each other, which offers additional evidence that these viral networks could be viewed as single, highly coupled modules (fig. S6). Because of these unusual topological features, viral networks should be more resistant to deliberate attacks than cellular networks, as both network size and characteristic path length remain more stable after the most highly connected nodes are removed (Fig. 1, C and D).

Whereas sequence and phylogenetic analyses identify a core set of genes conserved in all herpesviruses, the KSHV interactome allowed us to determine a core set of interactions conserved in all herpesviruses. Using the recip-
Comparison of network parameters of cellular and viral protein interaction networks. The table indicates key parameters of eight viral and cellular networks (all analyses were done using the SLCC of the respective network). Among 123 nonredundant KSHV protein interactions, 8 are self-interactions and the SLCC consists of 115 edges; among 173 nonredundant VZV protein interactions, 13 are self-interactions and one edge is isolated, leaving the SLCC with 159 edges. Only 1 of the 123 interactions in KSHV and 10 of the 173 in VZV were detected bidirectionally, and 1 interaction in KSHV and 13 in VZV were redundantly detected by distinct fragments of the same proteins. Interactions detected only in one direction are a common phenomenon in two-hybrid assays and most likely are due to steric hindrance of either bait or prey fusion proteins. The data sets were derived from the following sources: vaccinia virus, from McCraith et al. (13); S. cerevisiae I, from Schwikowski et al. (9); S. cerevisiae II, from the Database of Interacting Proteins (DIP; October 2004 release); H. sapiens I, from Rual et al. (8); H. sapiens II, from Stelzl et al. (7); and H. sapiens (predicted), from Lehner and Fraser (12). The table includes the number of nodes and edges in the SLCC; the average node degree (i.e., number of neighbors); the power coefficient $\gamma$ and its $P$ value (the slope and its significance under linear regression) as fitted by a power-law degree distribution (“scale-free” property); the characteristic path length and diameter, as well as the clustering coefficient and its fold enrichment over comparable random networks (“small-world” property). For each real network, a corresponding ER (Erdős-Rényi) randomization has the same number of nodes and edges, and an ES randomization, generated through an edge-swapping algorithm, also has the same degree distribution. The fold enrichments shown are over the theoretical clustering coefficient under the ER model and the median clustering coefficient of 1000 ES randomizations, respectively (see supporting online material). Note that the network parameters of the two yeast networks are surprisingly stable, although a large number of interactions have been included additionally into the DIP database in comparison to the initial data set generated by Schwikowski and colleagues (9). The recently reported H. sapiens networks have rather low levels of local clustering, which was discussed as being caused by their incompleteness (8). NA, not applicable owing to the low number of edges.

Table 1.

<table>
<thead>
<tr>
<th>Network parameters</th>
<th>KSHV</th>
<th>VZV</th>
<th>Vaccinia virus</th>
<th>S. cerevisiae I</th>
<th>S. cerevisiae II</th>
<th>H. sapiens I</th>
<th>H. sapiens II</th>
<th>H. sapiens (predicted)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Nodes</td>
<td>50</td>
<td>55</td>
<td>7</td>
<td>1,548</td>
<td>2,397</td>
<td>1,307</td>
<td>1,598</td>
<td>3,169</td>
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<tr>
<td>Edges</td>
<td>115</td>
<td>159</td>
<td>6</td>
<td>2,358</td>
<td>6,101</td>
<td>2,483</td>
<td>3,072</td>
<td>10,636</td>
</tr>
<tr>
<td>Average degree</td>
<td>4.60</td>
<td>5.78</td>
<td>1.71</td>
<td>3.05</td>
<td>5.09</td>
<td>3.80</td>
<td>3.84</td>
<td>6.71</td>
</tr>
<tr>
<td>Power coefficient</td>
<td>0.95</td>
<td>0.78</td>
<td>NA</td>
<td>2.14</td>
<td>2.01</td>
<td>1.54</td>
<td>1.66</td>
<td>1.81</td>
</tr>
<tr>
<td>$P$ value</td>
<td>$1.2 \times 10^{-4}$</td>
<td>$1.1 \times 10^{-4}$</td>
<td>NA</td>
<td>$3.6 \times 10^{-11}$</td>
<td>$7.7 \times 10^{-23}$</td>
<td>$1.2 \times 10^{-20}$</td>
<td>$5.8 \times 10^{-25}$</td>
<td>$1.4 \times 10^{-30}$</td>
</tr>
<tr>
<td>Characteristic path length</td>
<td>2.84</td>
<td>2.34</td>
<td>NA</td>
<td>7.28</td>
<td>5.10</td>
<td>4.36</td>
<td>4.85</td>
<td>6.40</td>
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<tr>
<td>Diameter</td>
<td>7</td>
<td>5</td>
<td>NA</td>
<td>23</td>
<td>13</td>
<td>12</td>
<td>13</td>
<td>20</td>
</tr>
<tr>
<td>Clustering coefficient</td>
<td>0.146</td>
<td>0.393</td>
<td>NA</td>
<td>0.213</td>
<td>0.296</td>
<td>0.060</td>
<td>0.012</td>
<td>0.186</td>
</tr>
<tr>
<td>Enrichment over ER</td>
<td>1.55</td>
<td>3.67</td>
<td>108.1</td>
<td>139.5</td>
<td>20.6</td>
<td>4.8</td>
<td>87.9</td>
<td>11.7</td>
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<tr>
<td>Enrichment over ES</td>
<td>0.76</td>
<td>1.01</td>
<td>29.2</td>
<td>29.5</td>
<td>0.92</td>
<td>0.28</td>
<td>1.2</td>
<td>1.14</td>
</tr>
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</table>

For each network, node degrees $k$ and their relative frequency (i.e., probability) are plotted on a bilogarithmic scale and fitted by linear regression. (C and D) Simulations of a deliberate attack on KSHV and yeast networks by removing their most highly connected nodes (in decreasing order). After each node is removed, the new network characteristic path length (average distance between any two nodes) and size (number of nodes) of the remaining single largest connected component (SLCC) are computed and plotted as a multiple or fraction of the original parameters. KSHV exhibits much higher attack tolerance, because the increase in path length and the decrease in network size are considerably smaller.
with viral interaction partners are not more conserved than are those without (fig. S7a), for KSHV proteins with interaction partners there was a significant correlation between the number of protein interactions and homology to the respective EBV ortholog (fig. S7b). When we compared the protein networks of KSHV and VZV, we found that only 9 of the 50 protein interactors in KSHV have orthologous interactors in VZV (Fig. 1A; table S1). Analysis of the KSHV network suggests that 19 interactions should exist between orthologous viral proteins in VZV, of which we could detect 5 in our Y2H screen (26%). This low result is not surprising, because the number of predicted interactions in HSV-1, mCMV, and EBV that were confirmed by Y2H was also considerably lower than the number confirmed by CoIP, indicating an inherent technical limitation of the Y2H system.

The network analyses of the KSHV and VZV interactomes revealed unique features of viral systems that also manifested themselves on the local level (fig. S8). Because we hypothesized that many of these features could be attributed to missing virus-host interactions, we modeled the interplay between viral and human protein networks. To date, only rather small subnets of the human interactome have been reported (7, 8). Unfortunately, only a few published human proteins targeted by KSHV lie within these reported human subnets and only a few virus-host interactions could be predicted on the basis of homology between KSHV and human proteins. However, using a prototypical human protein interaction network [derived from high-confidence interactions in S. cerevisiae, Caenorhabditis elegans, and Drosophila melanogaster (9–11)] that is considerably larger than the experimental subnets, we were able to efficiently connect the viral and the human networks by predicting interactions between KSHV and human proteins if both proteins have known interacting orthologs in either S. cerevisiae, C. elegans, or D. melanogaster (12). By this approach, we found 20 predicted interactions between 8 KSHV and 20 human proteins that are connected within the human network. Nineteen of these 20 virus-host interactions were tested by CoIP and an unexpectedly large percentage (13 out of 19, or 68%) could be confirmed (table S9). Although published viral-host interactions tend to involve genes or interactions specific to human or higher eukaryotes (because most human targets have no orthologs or orthologous interactions in the three lower-eukaryotic model organisms), our predicted viral-host interactions involve genes and interactions conserved from yeast to human and hence might reflect more general host-interacting mechanisms.

Using the predicted KSHV-human interactions, we were able to merge the two interactomes with each other (Fig. 2, A and B). The topology of the KSHV network changed

![Fig. 2. Interplay between the KSHV and a predicted high-confidence human network. (A) Global view of the interplay between the KSHV and a predicted high-confidence human interaction network consisting of 10,636 edges among 3169 nodes. Viral proteins are depicted as red nodes, cellular interacting proteins (level 1 and 2) as blue nodes, and cellular proteins (level >2) as gray nodes. Interactions between viral proteins are depicted as red edges, those between viral and cellular proteins as green edges, and those between cellular level 1 and 2 proteins as blue edges. (B) Local view of the combined KSHV and human interaction networks (level ≤2). (C) The combined viral-host network has a bigger and statistically more significant power coefficient and thus adopts scale-free features. The combined virus-host network was compared to 1000 random networks, which were generated by rewiring fixed virus interactors to swapped cellular proteins with the same degree as the actual target. The power coefficient and the power-law fit of the predicted (red triangle) and 1000 random (blue circles) KSHV-human networks are indicated on the x axis and the y axis, respectively. Among 1000 random networks, 65 have a higher power coefficient, 19 a more significant power-law fit, and only 8 a higher power coefficient that is more significant (empirical P value <0.01).]
Magnetosomes Are Cell Membrane Invaginations Organized by the Actin-Like Protein MamK

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Magnetosomes are membranous bacterial organelles sharing many features of eukaryotic organelles. Using electron cryotomography, we found that magnetosomes are invaginations of the cell membrane flanked by a network of cytoskeletal filaments. The filaments appeared to be composed of MamK, a homolog of the bacterial actin-like protein MreB, which formed filaments in vivo. In a mamK deletion strain, the magnetosome-associated cytoskeleton was absent and individual magnetosomes were no longer organized into chains. Thus, it seems that prokaryotes can use cytoskeletal filaments to position organelles within the cell.

Prokaryotes are highly organized cells with many ultrastructural similarities to eukaryotic cells (1). In addition to a dynamic cytoskeleton composed of homologs of actin, tubulin, and intermediate filaments (2), some prokaryotes also possess intracellular membranous organelles (3–6). The magnetosome chains of magnetotactic bacteria are among the best studied examples of membranous prokaryotic organelles. Magnetoosome chains contain 15 to 20 ~50-nm magnetite crystals that together act like the needle of a compass to orient magnetotactic bacteria in geomagnetic fields, thereby simplifying their search for their preferred microaerophilic environments (3). The distinct properties of magnetosomal magnetite crystals have drawn attention to their potential use in biotechnology, bioremediation, and geobiology (3). Furthermore, the cell biological characteristics of magnetosomes make them an ideal model system for the study of organelle biology in prokaryotes. Each magnetite crystal within a magnetosome is surrounded by a lipid bilayer, and specific soluble and transmembrane proteins are sorted to the magnetosome membrane (7, 8).

Electron cryotomography (ECT) is an emerging technology that can reveal the three-dimensional (3D) ultrastructure of small, intact prokaryotic cells in a near-native, “frozen-hydrated” state (9, 10). Here, we used ECT to define the precise physical characteristics of the magnetosome chain and to identify the molecular factors controlling its subcellular positioning and organization.

To image magnetotactic bacteria by ECT, exponential-phase cultures of *Magnetospirillum magnetica*um sp. AMB-1 were plunge-frozen and imaged in a transmission electron cryomicroscope. Dual-axis image tilt series were recorded (11), from which 3D reconstructions of individual cells were calculated (Movie S1). A slice through the middle of one typical reconstruction (Fig. 1A and Movie S2) clearly showed the inner and outer membranes and the peptidoglycan layer. The cytoplasm was full of ribosomes, recognized as the numerous, dense, ~25-nm particles. Large, approximately 10

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