Intrinsic Disorder Mediates Hepatitis C Virus Core – Host Cell Protein Interactions

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SUPPLEMENTARY MATERIAL

Introduction

The Supplementary Material includes additional discussions of the current status of computational methods for identifying IDR and MoRF and the results of the phylogentic analysis of the HCV MoRFs. We also include an expanded methods section, a table that summarizes the functions of the human proteins that interacted with HCV Core in this study, a list of primers used to clone HCV Core and NS5A fragments, and Supplementary Figure 1.

Co*mputational algorithms for IDR and MoRF identification –* Efforts to assess the prevalence and function of intrinsic disorder and MoRFs within proteomes have yielded over 50 bioinformatic predictors of disorder (reviewed in $¹$) and several MoRF predictors $²⁻⁵$. The earlier predictors $²$ focused on MoRFs</sup></sup></sup> that tend to form helices upon binding (called $α$ -MoRFs). However, MoRFs of various secondary structures and even combinations of secondary structures have been observed ⁶. General predictors have been developed to identify sequences that form irregular or "random coil" forms upon binding as well as those that form helix, sheet, or combinations thereof 5 ⁷. Other predictors of disorder $\frac{1}{2}$ and protein binding sites have also been developed, including several others that use linear sequence motifs for identifying binding sites ⁷⁻¹⁰. These motif-based methods do not explicitly consider structure of the conserved motifs, but these linear binding motifs 11 are mostly located in regions of predicted disorder $12,13$.

To identify IDRs and MoRFs within the HCV polyprotein, we employed the Predictors of Naturally Disordered Regions (PONDR®), PONDR®-VLXT¹⁴ and the "meta-predictor", PONDR®-FIT¹⁵. The PONDR®-FIT meta-predictor combines three predictors from the PONDR® family with three additional predictors of disorder, namely IUPred 16 , FoldIndex 17 , and TOP-IDP 18 . This meta-predictor performs better than any of the individual predictors, providing one of the best predictions of disorder across a protein sequence ¹⁵. PONDR®-VLXT yields better sensitivity in predicting alpha-helical propensity and, in combination with the α -MoRF-Pred computational tool, can identify putative α -MoRFs⁴. Although several additional predictors have been developed to identify likely MoRFs within longer regions of disorder, we chose this particular predictor because of previous success with this approach.

Despite the abundance of computational tools, very few studies have integrated binding site predictions to help identify and characterize protein-protein interaction binding interfaces. When PONDR®-VLXT disorder prediction was applied to the RNA degradosome, the C-terminal half of the protein was strongly predicted to be disordered but with three different short predicted regions of structure, called "regions of increased structural propensity." These three regions were each shown to bind to different protein partners (12). Consistent with the computational prediction, one of these short regions was partially helical in the X-ray crystal structure of the complex with enolase ¹⁹. In a second study, computational analyses of amino acid sequences of the proteins involved in protein-protein interaction networks were used to identify likely linear motifs ²⁰. Two previously untested linear motifs were then shown experimentally to bind with their indicated partners 20 . Neither of these prior studies carried out experiments to test whether the same MoRF or linear motif could bind to more than one partner as shown herein.

Core MoRFs are undergoing purifying selection – Observing stronger negative, or purifying, selection within the predicted MoRFs would provide independent evidence of a functional role for these features in HCV biology. To assess the levels of purifying selection acting on specific partitions of the HCV Core protein, we used Bayesian phylogenetic inference to estimate the site-specific ratios of non-synonymous and synonymous substitution rates, dN/dS, from Core coding sequences from 86 human HCV isolates representing genotypes 1-7. Each sequence was divided into five partitions corresponding to the two MoRFs, the non-MoRF IDR, the entire IDR (domain I), and the structured C-terminal region (domain II). Mean dN/dS values were calculated for each sequence partition from each sampled tree and the

distributions of the means were used to determine the 95 and 99% Bayesian confidence intervals (Table 1). MoRF1 and 2 had significantly lower dN/dS values than the surrounding IDR (P<0.01 for both), indicating that they have undergone stronger purifying selection (Fig. 2C). This analysis also revealed that the structured C-terminus of Core, which mediates the interaction with cytoplasmic lipid droplets (residues 118-191 in JFH1 Core), has undergone stronger purifying selection than the Core IDR (residues 1-117 in HCV 2a JFH1 Core).

The picture for NS5A is more complicated than for Core. Overall, there is greater sequence variation in the NS5A IDRs, with many insertions and deletions that complicate the alignment (Fig. 2B). As with Core, 76 NS5A sequences from all 7 HCV genotypes were divided into MoRF and non-MoRF partitions. Estimates of the dN/dS ratios for the NS5A MoRFs showed that MoRFs 1 and 3 were significantly lower than the flanking IDR (P<0.01 for both) (Fig. 2D and Table 2). The dN/dS estimates for NS5A MoRF2 and 5 suggested that they are undergoing purifying selection (P<0.05 for both) but were not significantly lower than the dN/dS ratio estimate for the flanking IDR. Surprisingly, the dN/dS estimates for the MoRF 4 partition indicated that it is undergoing diversifying selection $(dN/dS > 1, P < 0.01)$.

Supplementary Methods

Yeast two-hybrid assays – Yeast two-hybrid library screens were performed as described in ²¹⁻²⁴. Briefly, MoRF-containing fragments of the HCV JFH1 polyprotein (UNIPROT: Q99IB8) were cloned into the yeast two-hybrid DNA-binding domain (DBD) plasmid pOBD2 by homologous recombination in the yeast strain R2HMet (*MAT*a *ura3-52 ade2-101 trp1-901 leu2-3,112 his3-200 met2*∆*::hisG gal4*∆ *gal80*∆) $24,25$. The primers used to generate the fragments are shown in Table 1. Expression of all DBD constructs generated in this study was confirmed by western blot as described in 26 . All constructs were verified by PCR and sequencing. Yeast two-hybrid assays were performed by mating yeast expressing the DBD fusion proteins with yeast strain BK100 (*MATa ura3-52 ade2-101 trp1-901 leu2-3,112 his3-200 gal4*∆ *gal80*∆ *GAL2-ADE2 LYS2::GAL1-HIS3 met2::GAL7-lacZ*, a derivative of PJ69-4A) 24,27,28 that contained a human gene fragment cloned into the activation domain (AD) plasmid pOAD.103^{21,29}. Diploid yeast were selected on synthetic dropout (SD) media lacking tryptophan and leucine (SD-TL), then plated on SD media lacking tryptophan, leucine, uracil, and histidine, plus 3-amino-1,2,4-triazole at concentrations empirically determined to inhibit background growth (Y2H selection media).

Split-luciferase assays – HCV Core fragments 1 and 2 were cloned into p424-BYDV-NFLuc, whereas each cellular binding partner of Core MoRF1 and MoRF2 were cloned into p424-BYDV-C-Fluc-FLAG $30.$ EGFP was cloned into both vectors and served as a negative control to determine the background in the assay for each fusion protein. Fusion proteins were expressed in TNT® SP6 High-Yield Wheat Germ Lysate (Promega), subjected to SDS PAGE and western blotting with anti-FLAG (Sigma, F1804) or anti-NFLuc (Santa Cruz Biotechnology, sc-57603) antibodies. Based on the abundance of the proteins on the western blots, the concentrations of the samples were normalized to the EGFP controls. For splitluciferase assays, MoRF1, MoRF2, and EGFP were combined with the cellular binding partners or EGFP as a negative control. After incubating 1 hour at 4°C, binding reactions were combined with luciferase substrate solution and luminescence was assessed. Luminescence for each reaction was compared to that of each control. For an interaction to be considered significant, the luminescent signal must be higher than both control reactions (NFLuc-Test + CFLuc-EGFP and NFLuc-EGFP + CFLuc-human protein) as determined by a one-tailed t-test (P<0.05).

Error-prone mutagenesis – Error-prone mutagenesis of the MoRF-containing fragments of the HCV Core protein were generated as described in $31,32$. Nucleotide analogs 8-oxo-2' deoxyguanosine (8-oxo-dGTP) and 6-(2-deoxy-b-D-ribofuranosyl)-3,4-dihydro-8H-pyrimido-[4,5-c][1,2]oxazin-7-one (dP) were added to standard GoTaq PCR reactions (Promega) at concentrations of 1 and 10uM. The mutagenic reactions were run for five cycles, after which the products were purified to remove unincorporated analogs and used as templates for another 25 cycles. PCR products were inserted into pOBD2 by homologous recombination.

Supplementary Table 1. The human binding partners of HCV Core. Description of the functions of Core-interacting proteins and any previously identified roles in HCV infection. Start and end refer to the amino acid positions of the fragments identified in the yeast two-hybrid screens.

Supplementary Fig. 1. Western analysis of split-luciferase fusion proteins. FLAG- and CFLuc tagged human proteins (A) and NFLuc-HCV Core (B) fusion proteins were in vitro translated in wheat germ extracts (Promega). Equal volumes of each sample were subjected to SDS PAGE followed by immunoblotting with anti-FLAG or anti-NFLuc antibodies. The concentrations of fusion proteins were adjusted based on the western blot results so that the test protein concentrations were equal to (or slightly less) than the GFP negative controls.

Supplementary Fig. 2. A potential inverted linear motif in Core MoRF2. Alignment of the reversed NPHV sequences with HCV MoRF2 yielded the consensus sequence GWAxWxLxP, which may represent an inverted protein-binding site in Core. Such sites, which have been termed retro-MoRFs, have been proposed 46, but to date very few such retro-MoRFs have been functionally characterized 47. **(**A) Block Logo and Web Logo of Core residues 89-100 (based on HCV JFH1 numbering) from HCV genotypes 1-6 (GTs) and 8 Hepacivirus isolates. The Block Logo (A, top) shows the frequency of specific peptide sequences, with the relative height of each peptide reflecting its frequency in the input alignment. In contrast, the Web Logo shows the conservation of residues at each position. (B) Block Logo and Web Logo of Core residues 89-100 with the same region from the Hepaciviruses inverted. The Web Logo of the HCV and the inverted Hepacivirus sequences revealed a conserved linear motif at position 92-100, GWAxWxLxP.

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