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LETTER TO THE EDITOR

Real-World Data on Antiviral Treatments for Hepatitis C Virus Infections: Can We Define Intention to Treat or Per Protocol Analyses?

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We read with interest a recent study that used real-world (i.e. observational) data from the German Hepatitis C-Registry.¹ The 12-week sustained virologic response (SVR) was compared between 8- and 12-week regimens of ledipasvir/sofosbuvir. The authors used classifications of intention to treat (ITT) and per protocol to define eligibility of patients for their analyses.¹ One classification of ITT defined eligibility as patients who completed treatment with either the 8- or 12-week ledipasvir/sofosbuvir regimen (SVR≈85%). The second classification of ITT defined eligibility as patients who initiated and completed either the 8- or 12-week ledipasvir/sofosbuvir regimen (SVR≈95%). Per protocol defined eligibility as patients who initiated and completed treatment, adhered to treatment throughout the duration, and had SVR status assessed 12 weeks post-treatment completion (SVR≈98%). Nevertheless, such use of the terms ITT and per protocol have undue popularity in real-world studies of direct-acting antivirals (DAAs).¹⁻⁴

ITT and per protocol are approaches to statistical analysis of randomized controlled trials (RCTs) and pertain to treatment status.⁵ Neither approach is used to define eligibility and neither approach is directly applicable to real-world studies. ITT involves analyzing outcomes for RCT participants based on the treatment to which they were randomized, regardless of adherence to the allocated treatment.⁵ ITT preserves the balance of known and unknown confounders between comparison groups (i.e. exchangeability), which is the key benefit of randomization for causal inference. In contrast, real-world studies have an inherent risk of confounding by indication, which no form of statistical adjustment can completely resolve.⁶ Per protocol involves analyzing outcomes for participants based on adherence with the allocated treatment, which addresses the issue of treatment misclassification.⁵ The potential reduction in treatment misclassification comes with the trade-off of breaking randomization; a per protocol analysis effectively converts the trial to a quasi-experimental study.⁵

The approach by Buggisch et al.¹ and others is incompatible with ITT or per protocol definitions, and raises serious concerns about overestimated SVR in real-world studies. To facilitate awareness of biasing mechanisms, *Figure 1* illustrates the four possible types of HCV-infected patients who initiated DAAs in any real-world study regardless of treatment duration. The distribution of these four patient types across regimens (e.g. 8- or 12-weeks) ultimately determines the observed SVR incidence. Given that we cannot rely on randomization to designate treatment status as in an RCT and the planned treatment duration (8 weeks or 12 weeks) was not recorded in the registry,¹ we must rely on exposure to treatment for eligibility. Treatment duration could have been modified based on an intermediate measure of response, which exacerbates the potential for confounding by indication. Nevertheless, we emphasize that all four patient types would be eligible for the analysis. Type 1 and 2 patients were followed through treatment completion and 12-week SVR assessment, and SVR was achieved by type 1 but not type 2 patients. Type 3 patients completed treatment, but the SVR status was unknown because of loss to follow-up (e.g. some barrier to care), whereas type 4 patients were lost to follow-up before completing treatment (e.g. side-effects or other reasons for discontinuation) and SVR status was also unknown. Buggisch et al.¹ excluded type 4 patients because of missing SVR status, but these patients were eligible albeit unlikely to achieve SVR. In addition, the main analyses (labeled “per protocol”) excluded type 3 patients because of missing SVR status despite SVR being possible but unknown. Such exclusion relies on the unrealistic assumption that excluded cases were missing completely at random.⁷ Non-random exclusion of patients based on outcome status leads to a selected population of patients who completed treatment and had a high probability of SVR. Even the lowest estimate of SVR reported in the study (85%) may be an overestimate.

Given well-known problems with complete case analysis,⁷ the challenge is how to handle patients with missing outcome data because of loss to follow-up. This challenge applies to RCTs and real-world studies, and no consensus has been established about the best approach.

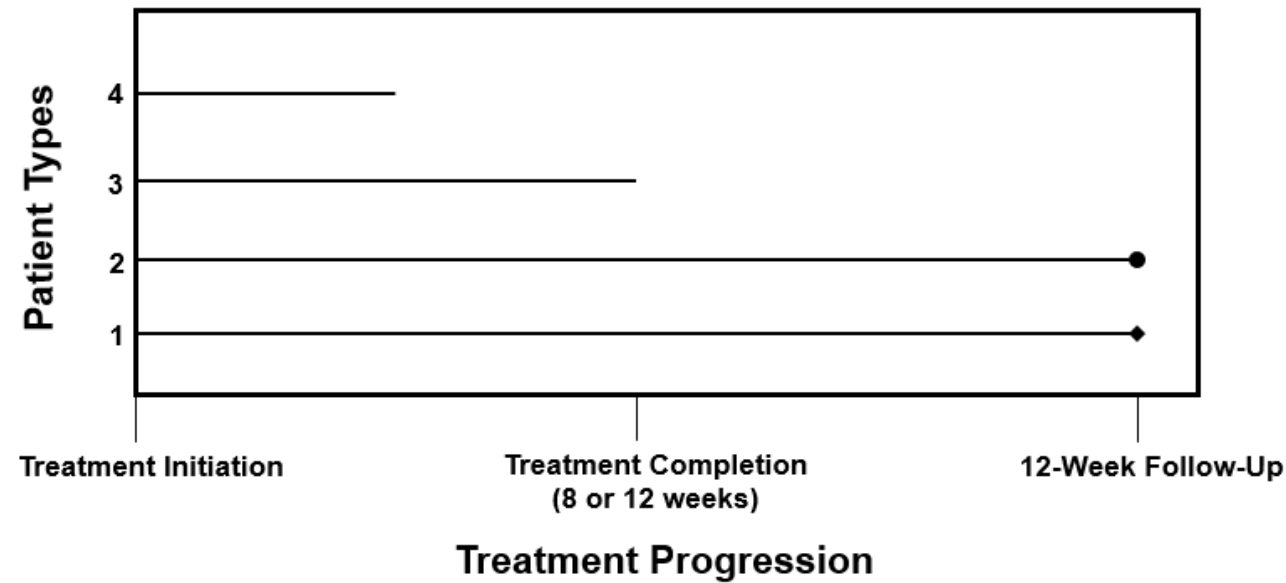
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4 A simple approach is to designate worst-case and best-case scenarios, where none of the
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6 individuals with missing SVR status would have achieved SVR or all of the individuals would
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8 have achieved SVR, respectively.⁸ The range of estimates based on these designations can be
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10 informative unless extensively missing SVR status is present.⁸ More sophisticated approaches
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12 include multiple imputation and inverse probability weighting, but these approaches are not
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14 necessarily superior in all scenarios.⁸⁻¹⁰
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17 We conclude that the interpretation of favorable response with 8- or 12-week treatment
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19 and the observed small differences between these regimens is problematic. Some limitations of
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21 real-world data cannot be overcome. Sensitivity analyses and cautious interpretation are
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23 encouraged.
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Figure 1. Distribution of patient types and follow-up in real-world studies of direct-acting antivirals for hepatitis C virus infections.^a



^aThe diamond at 12-week follow-up for patient type 1 represents sustained virologic response (SVR) and the circle for type 2 represents no SVR. Patient types without these symbols indicate loss to follow-up.

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