A Case Study Using KEYNOTE-010 to Compare and Evaluate Long-Term Survival Estimates from Two Classes of Piecewise Models

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Objective

To compare the accuracy of long-term survival estimates from a recently developed piecewise exponential model with piecewise models using Kaplan-Meier data adjoined to a parametric tail.

Background

- Piecewise survival models have been suggested as a flexible alternative to standard parametric models for modeling complex hazard profiles, such as those commonly associated with immune-oncology (IO) therapies.
- One common class of piecewise model uses Kaplan-Meier (KM) data adjoined to a parametric tail (PW-KM) with the cut-point(s) defined through visual inspection of hazard plots for distinct changes or inflection points in the hazard profile.
- However, there are no definitive rules for the selection of the 'best' cut-point(s).¹
- A method to statistically estimate the cut-point(s) has recently been developed by Cooney et al. (2023) using a piecewise exponential (PWE) model where a constant hazard is assumed within each segment, and separate exponential parametric models are fitted to each segment.² The final segment serves as the long-term extrapolation.
- Both model classes are displayed conceptually in Figure 1.
- This case study was motivated by the need for continued research into the comparison of method performance so survival models can be appropriately selected and lead to more robust long-term survival estimates.

Methods

KEYNOTE-010

 KEYNOTE-010 investigated pembrolizumab, an IO therapy for previously treated, programmed death-ligand 1-positive advanced non-small cell lung cancer.

Data Generation

- Published overall survival KM curves from two data cut-offs (DCOs) were digitized:
- First DCO: median follow-up 31.0 months.³
- Second DCO: median follow-up 67.4 months.⁴
- Pseudo individual patient data (IPD) were generated from each DCO using the algorithm described by Guyot et al. (2012).⁵

Model Fitting

- All models were fitted to pseudo IPD derived from the 31.0-month DCO.
- The PWE model identified a single cut-point at 13 months, and separate exponential models were fitted to each segment.
- For the PW-KM model, five months was selected as a cut-point through visual inspection of hazard plots, as the smoothed hazard plot indicated that at five months, the hazard became monotonically decreasing. A second PW-KM model with a 13-month cut-point was additionally specified to facilitate comparison with the PWE model.
- For the PW-KM models, from the cut-points onward, the six standard parametric tails were fitted to the remaining KM data and adjoined to the KM curves.

Long-Term Survival Comparison

- To evaluate accuracy of the survival models in estimating long-term survival, restricted mean survival time (RMST) at maximum follow-up from the KEYNOTE-010 67.4-month DCO was compared to the estimated 67.4-month RMST based on extrapolation of the 31.0-month DCO for each model.
- Accuracy was assessed using absolute percentage difference in RMST calculated as:

RMST_{estimated} – RMST_{actual} RMSTactual

Results

- RMST from the 67.4-month DCO at maximum follow-up was 22.66 months.
- The 67.4-month DCO KM curve and the survival extrapolations for each model are shown graphically in **Figure 2**.
- Estimated RMST from the PWE model was 21.76 months, and average estimated RMSTs across the parametric tails for the 5- and 13-month PW-KM models were 22.46 (range: 20.52–23.79) and 22.53 (range: 21.71–24.07), respectively.
- Long term survival estimate accuracy results are presented in **Figure 3**.
- The most accurate model was the 5-month PW-KM model with a GenGamma tail (absolute percentage difference in RMST: 0.64%) followed by the 13-month PW-KM model with a Gompertz tail (absolute percentage difference in RMST: 1.11%) and the 5-month PW-KM with a LogLogistic tail (absolute percentage difference in RMST: 1.14%).
- The PWE model (absolute percentage difference in RMST: 3.96%) was ranked 7/13 in terms of model accuracy.

Conclusion

- In this case study, the PWE model did not perform better than the PW-KM models in estimating long-term survival, on average.
- In terms of absolute percentage difference, the PWE, 5-month PW-KM, and 13-month PW-KM models all performed similarly, with the 13-month PW-KM performing best by a slight margin.
- Although the cut-point was identified objectively for the PWE model, specifying the model to use constant hazards may limit its accuracy particularly for the final segment where the exponential model cannot reflect a decreasing hazard in the KM tail.
- Biological plausibility and hazard plots should be assessed to validate the required hazard assumptions.
- Nonetheless, the fact that the PW-KM using the cut-point identified by the PWE model performed best suggests that further research into objectively identifying cut-points should be explored.
- Limitations of this research are that it was a single case study and may not be generalizable, and that 'accuracy' was measured against realized RMST from long-term observed KM data that are themselves uncertain.

Abbreviations: DCO: data cut-off; IO: immune-oncology; IPD: individual patient data; KM: Kaplan-Meier; PWE: piecewise exponential; PW-KM: piecewise model using Kaplan-Meier data adjoined to a parametric tail; RMST: restricted mean survival time. References: ¹Liu B. et al. Presented at ISPOR, 15–18 May 2022. Washington DC, USA. EE111; ²Cooney P. et al. J Thorac Oncol 2021;16(10):1718–1732; ⁵Guyot P. et al. BMC Med Res Methodol 2012;12:9. Acknowledgments: The author thanks Freya Turner, Costello Medical, for graphic design assistance, Hannah Harrington, Costello Medical, for assistance generating pseudo IPD, and Molly Atkinson, Costello Medical, for review and editorial assistance in the preparation of this poster.









A. 5-month cut-point PW-KM 0.9 -0.8 -0.7 -0.6 -0.5 -0.4 -0.3 -

0.9 -0.8 -0.7 -0.6 -0.5 -0.4 -0.3 ഗ് _{0.2} .

0.9 -0.8 -0.7 -0.6 -0.5 -- 4.0 **Xal** 0.3 -0.2

FIGURE 1

FIGURE 2











MSR24



FIGURE 3

Long-term survival prediction accuracy

Absolute percentage difference in RMST 🛛 — Mean absolute percentage difference in RMST



