

# Automatic Planning Results Using a Novel Dose Prediction Tool

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## Deciding on the underlying technology

Compared trade-offs between various techniques

Knowledge-based	Multi-criteria optimization	Progressive Optimization
<ul style="list-style-type: none"> <li>Minimal user input required</li> <li>Adapts to planning trade-offs</li> </ul> <p><b>BUT</b></p> <ul style="list-style-type: none"> <li>Dependent on a knowledge base</li> <li>Not flexible to inter-physician variability</li> <li>Affected by variations in contouring</li> <li>Does not address new knowledge on toxicity endpoints</li> </ul>	<ul style="list-style-type: none"> <li>Provides trade-off analysis with interactive graphical interface</li> </ul> <p><b>BUT</b></p> <ul style="list-style-type: none"> <li>Requires most physician time</li> <li>Plan quality degrades in conversion</li> <li>Does not lend to Standardization</li> </ul>	<ul style="list-style-type: none"> <li>Minimal user input</li> <li>Adaptable to protocol changes</li> <li>No knowledge-based required</li> <li>Dosimetric drivers not limited to DVH parameters</li> </ul> <p><b>BUT</b></p> <ul style="list-style-type: none"> <li>No on-the-fly trade-off analysis</li> <li>No historical information</li> </ul>

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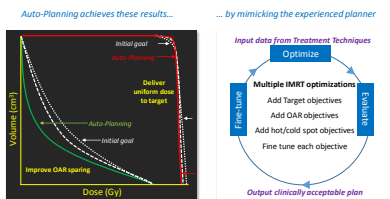
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## Progressive optimization algorithm

Drives target coverage and sparing to the limits




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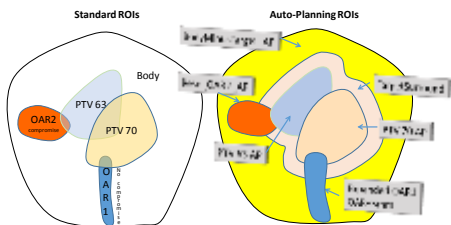
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Auto-Planning ROIs




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Progressive optimization algorithm  
Drives target coverage and sparing to the limits



Even with generic inputs, Auto-Planning pushes beyond what was requested

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Progressive optimization algorithm  
Validated Through Peer-Reviewed Research

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V. CONCLUSIONS

Comparison of autoplans and previous delivered clinical plans showed only small dosimetric differences in target coverage, but significant reduction in dose to OAR for the autoplans. The blinded clinical evaluation of the plans showed that, for 94% of the evaluations, the autoplans were similar to or better than the clinical plans. Auto-Planning software will, therefore, be able to reduce the manual time spent per treatment plan since the most of the plans could potentially be used clinically without further optimization. Perhaps more importantly, Auto-Planning could be used as a high quality starting point for further plan optimization. This could increase the overall quality of the treatments and reduce the interobserver variation present in manually created treatment plans.

Copyright 2016, PMA, Inc. All rights reserved. A representation of plan quality was achieved and the efficacy was improved for plan optimization and treatment.

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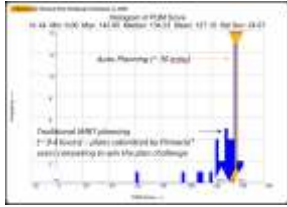
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### Auto-Planning at the Pinnacle<sup>3</sup> Plan Challenge Better results than the median at a fraction of the time

- Plan Challenge commissioned by Philips in 2013 through ROR
- Blue bars indicate submitted scores by users generating manual plans
- Yellow triangle indicates score achieved by Auto-Planning – well above the mean score




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Can we improve Auto-Planning results with patient-specific, *personalized*, inputs?

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### PlanIQ Feasibility Providing achievable sparing goals

- Clinical Goals Feasibility
- Distinguish between achievable and unachievable clinical goals
  - Efficiently modify clinical goals prior to the planning process
- Feasibility DVH<sup>TM</sup>
- Patient specific DVH for targets and OAR with feasibility bands
  - Optimizes treatment plan goals




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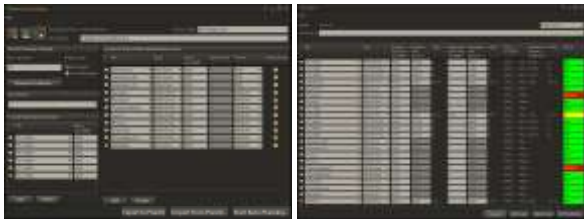
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### Feasibility driven results

Drives target coverage and sparing to the limits



- Auto-Planning still pushes many structures below feasibility numbers – SpinalCord, Brainstem\_03, Parotid\_L, Esophagus\_Up
- Other structures (Larynx, Pharynx, Oral Cavity) well within protocol guidelines, but not matching feasibility
- Great results, but can we make them better?

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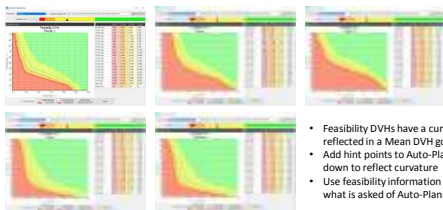
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### Progressive optimization algorithm

Drives target coverage and sparing to the limits



- Feasibility DVHs have a curvature that is not reflected in a Mean DVH goal
- Add hint points to Auto-Planning to have it drive down to reflect curvature
- Use feasibility information to drive priorities and what is asked of Auto-Planning

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### Feasibility Aiding Clinical Decision Making



- Add Max DVH Points to reflect FDVH curvature
- Change priorities based on Feasibility
- Remove structures that have no chance of being spared – in this case Submandibula\_R already removed

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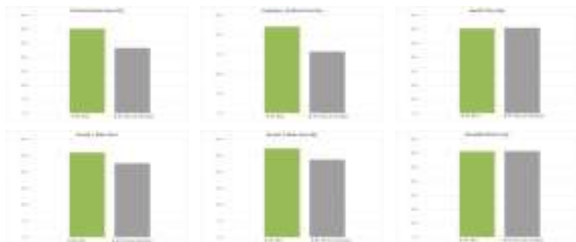
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Average results over 10 cases



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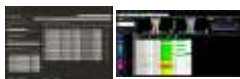
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Automatic Planning with Dose Prediction  
Conclusions



- Auto-Planning alone has been shown through peer-reviewed research to improve plan quality
- Personalizing Auto-Planning inputs through PlanIQ Feasibility based dose objectives can further improve OAR sparing
- PlanIQ Feasibility can improve up-front clinical decision making prior to plan creation by
  - Preventing optimization against unachievable goals
  - Providing more complete information on achievable goals based on patient geometry
  - Improving goal priority setting (high, medium, low) prior to planning based on patient geometry

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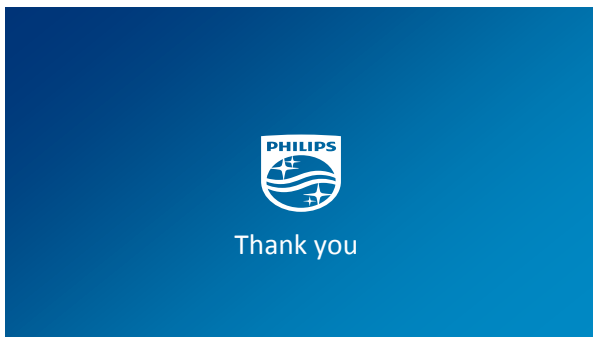
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