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Title: Transplant rejection and tolerance – advancing the field through integration of computational and experimental investigation

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1    **Editorial: Transplant rejection and tolerance – advancing the field**  
2    **through integration of computational and experimental**  
3    **investigation**

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31 Seventy years after the first proof of concept that the immune system can be trained to accept  
32 transplanted tissues via induction of immune tolerance, we are still waiting for a clinical  
33 approach that could be used routinely in transplant patients. Transplantation is a life-saving  
34 surgical procedure that is still only successful when paired with life-long administration of  
35 immunosuppressive drugs. However, the debilitating side effects of the long-term use of these  
36 drugs, together with their incomplete control of the immune system, compromise the quality of  
37 life and survival of transplant recipients. Thus, there is a strong push to find new therapeutic  
38 strategies that promote indefinite acceptance of a transplanted tissue without compromising the  
39 effectiveness of the patient's immune system. Although many exciting ideas have been explored,  
40 none of the resulting strategies have been successfully converted into a widely applicable  
41 therapeutic approach.

42  
43 Our knowledge of the complex immunological processes leading to transplant rejection  
44 continues to grow, and our understanding of the limitations associated with experimental models  
45 deepens. There is a great opportunity to foster a different approach to identify novel  
46 interventions. New tools of genomics, proteomics, and metabolomics are being implemented in  
47 powerful analyses that promise the development of better and safer personalized treatments. In  
48 parallel, theoretical modeling is slowly but progressively being welcomed among  
49 experimentalists due to its ability to unravel relevant mechanisms of complex systems and  
50 generate new hypotheses (1). The successful employment of these promising tools requires  
51 effective communication and collaboration among immunologists, data-driven modelers, and  
52 system biologists.

53  
54 This Research Topic provides a venue for stimulating these interdisciplinary conversations in the  
55 context of transplantation. The articles collected under this Research Topic introduce new  
56 theoretical and experimental studies that describe novel techniques and methods for  
57 understanding the interactions between the immune response and transplants and for establishing  
58 more effective strategies of diagnosis and intervention that will promote transplant tolerance.  
59 The contributions of this Research Topic can be divided into two main groups according to the  
60 approaches they implement: (i) big data and bioinformatics and (ii) mechanistic and equation-  
61 based models of rejection.

62  
63 To identify correlations and sensitivities from large data sets, various statistical methods and  
64 bioinformatics approaches are needed. Wang and Sarwal (2) offer a concise review of the current  
65 uses and advances in statistical approaches and high-dimensional data applications for  
66 identifying possible transplant biomarkers. Identifying markers of injury, causative markers, and  
67 predictive markers is key for monitoring, managing patients, and identifying the re-purposing  
68 potential of existing drugs. Mastoridis et al. (3) review current techniques (transcriptomic  
69 technologies) and propose future ideas for identifying biomarkers predictive of tolerance in the  
70 context of liver transplantation. They also explore how this knowledge could offer great insight  
71 into studying tolerance to other organs. In their perspective article, Stegall and Borrows (4) argue  
72 that more accurate and mechanistic mathematical models can be designed to predict (renal)  
73 allograft loss or chronic injury, but they note that this will require access to more detailed  
74 molecular, histologic, and serologic data. Mechanistic studies conducted in parallel to focused

75 clinical trials also would be tremendously useful for understanding why grafts fail and for  
76 designing tailored intervention.

77 Several statistical methods are applied to transplant data in articles of this collection to identify  
78 key biomarkers. Pike et al. (5) used principle component analysis and other tools to analyze a  
79 large set of T cell immunophenotyping data before and after renal transplantation. They  
80 discovered that pre-transplant frequency of **programmed** death 1 (PD-1) expressing T cell subsets  
81 stratifies patients at risk of developing rejection episodes. In a study of kidney transplants,  
82 Kadota et al. (6) used various statistical algorithms to analyze the transcriptome of allograft  
83 biopsies and showed that histological classification of T cell mediated rejection contains multiple  
84 subtypes of rejection amenable to more personalized treatments. When studying the  
85 inflammatory response associated with ischemic injury, Starzl et al. (7) combined principal  
86 component analysis and a regression approach to discover a cytokine-based signature to define  
87 the type and severity of the inflammatory response.  
88

**Deleted:** protein

89 In transplant modeling, identifying the key players and interactions between transplants and the  
90 immune system is critical to understanding the pathway to rejection or tolerance. An agent-based  
91 model presented by An (8) provides a dynamic and mechanistic understanding of transplant  
92 immunology so that control strategies to induce tolerance can be built. Arciero et al. (9) provide  
93 one of the first comprehensive mathematical models of mouse heart transplant rejection. This  
94 ordinary differential equation-based model tracks innate and adaptive immunity and provides  
95 important suggestions of new investigations to improve the understanding of rejection. Day et al.  
96 (10) present an ordinary differential equation model focused on the inflammatory response to  
97 surgical and ischemia/reperfusion injury. The model predicts specific conditions that lead to  
98 tolerance and others that lead to an exaggerated rejection response. Best et al. (11) use a  
99 computational model of T cell repertoire development to examine self/non-self discrimination  
100 when incorporating features of cross-reactivity and T cell cooperativity. The resulting dynamic  
101 state of tolerance suggests specific opportunities for therapeutic intervention to achieve long-  
102 term tolerance.  
103

104 Overall, all of the contributions to this Research Topic highlight the still largely untapped  
105 potential of integrating data-driven and mechanistic modeling into the “ordinary” experimental  
106 scientific approach to address key questions of transplant immunology in academic settings. As  
107 noted at a recent workshop of computational and experimental immunologists convened by the  
108 NIAID (12), there is still a broad divergence among researchers on how to approach fundamental  
109 immunological questions. This separation between modelers and experimentalists is even deeper  
110 in transplant immunology. However, all researchers share the common goal of improving the life  
111 of transplanted patients by understanding how to predict the behavior of immunological  
112 responses underlying graft rejection and failure. Despite the continuous growth of technological  
113 advances, it is still difficult to predict how a certain molecular or cellular intervention will affect  
114 the behavior of the entire system over time. This could be achieved, however, by properly  
115 integrating experimentation, data-driven modeling, and mechanistic modeling to test non-  
116 intuitive conditions impractical to explore using experimentation alone. The close collaboration  
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119 between experimentalists and modelers necessary to reach this result requires a novel component  
120 of formal training of each part that will lead to productive communication and work integration.  
121 This Research Topic encourages the research community to embrace and implement this  
122 approach and witness exciting new discoveries that will ultimately benefit the patient population.

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