Transplant immunosuppressant therapies induce unique functional changes in lymph node and re-shape gut immune homeostasis

Long Wu[1,2,+], Allison Kensiski[2,+], Samuel J Gavzy[1,2], Yang Song[3], Hnin Wai Lwin[3], Michael France[3,4], Dejun Kong[2], Lushen Li[1,2], Ram Lakhan[1,2], Vikas Saxena[1,2], Wenji Piao[1,2], Marina W. Shirkey[2], Valeria Mas[1], Brendan Lohmsr[1,2], Jonathan S Bromberg[1,2,4]*, Bing Ma[3,4]*

[1]University of Maryland School of Medicine, Department of Surgery, Baltimore, MD, USA

[2]University of Maryland School of Medicine, Center for Vascular and Inflammatory Diseases, Baltimore, MD, USA

[3]University of Maryland School of Medicine, Institute for Genome Sciences, Baltimore, MD, USA

[4]Department of Microbiology and Immunology, University of Maryland School of Medicine, Baltimore, MD, USA

[+]Authors contributed equally to this work

*Corresponding Authors:

Jonathan S. Bromberg, MD, PhD

jbromberg@som.umaryland.edu

Bing Ma, PhD

bma@som.umaryland.edu

Abstract

Current immunosuppressants effectively suppress adaptive and innate immune

responses, but their broad, antigen-non-specific effects often result in significant

complications. Here, we conducted a systematic investigation of drug effects on gut

microbiome, metabolic pathways, lymphoid architecture and lymphocyte trafficking of

four major immunosuppressants classes including tacrolimus, prednisone,

mycophenolate mofetil (MMF), and Fingolimod (FTY). We show that

immunosuppressants induced progressive alterations in the gut microbiome and

metabolic pathways, where initial drug-specific effects converged by day 30 into a

common dysbiotic state enriched in Muribaculaceae. Gut transcriptome analyses

revealed minimal overlap in differentially expressed genes between early and late time

points, indicating transient early effects and marked late-stage changes induced by

these drugs. These microbiome and metabolic shifts were accompanied by time-

dependent changes in lymph node (LN) organization and cellular composition,

transitioning between pro-tolerogenic and pro-inflammatory states. Under alloantigen-

stimulated conditions, MMF and FTY suppressed inflammation through dual regulation

of regulatory T cells and LN remodeling, mediated through LN stromal cell-derived

laminins. Together, our findings highlight previously underappreciated temporal

dynamics in immunosuppressant mechanisms, linking immune tissue organization, gut

microbiome and intestinal responses to long-term immunosuppression and metabolic

complications in transplant patients.

Introduction

Prevention of solid organ transplant rejection necessitates lifelong, multimodal immunosuppression to dampen both adaptive and innate alloreactive immunity. Despite improved short-term outcomes, long-term graft survival remains suboptimal due to chronic rejection and a range of adverse events such as metabolic complications, infections, neoplasia, and organ-specific toxicities [1-8]. Current immunosuppressants target specific immune pathways but are limited by their antigen-non-specific nature, broadly suppressing the immune system and causing off-target effects that impact non-immune cells[9, 10]. This dilemma between necessary immunosuppression and its imprecise, system-wide effects represents a critical challenge for improving transplant outcomes.

A comprehensive understanding of the immune suppressant drug effects is crucial to optimize therapeutic regimens and minimize adverse events while maintaining effective prevention of graft rejection. While most mechanistic research on immunosuppressants has focused on their effects on T cells, B cells, and dendritic cells, there is limited understanding of how these drugs impact secondary lymphoid organs (SLOs), particularly lymph nodes (LNs), which are key regulators of local and systemic immunity, directing responses toward either immune tolerance or activation. The architecture of LNs is maintained by LN stromal cells (LNSCs) networks, including fibroblastic reticular cells (FRCs), lymphatic endothelial cells (LECs), and blood endothelial cells (BECs). These LNSCs are critical for regulating the positioning and interactions of lymphocytes with antigen-presenting cells like DCs through the coordinated action of chemokines,

cytokines and stromal fibers [11-13]. FRC-derived laminins, in particular, influence

global immune states by regulating the balance between tolerance and immunity [14,

15]. Specifically, a high laminin α4:α5 ratio (La4:La5) within the T cell cortex of the LN

cortical ridge (CR) and around high endothelial venules (HEVs) promotes a pro-

tolerogenic environment, while a low La4:La5 fosters inflammation and immunity.

Additionally, LN structure and FRC function can be disrupted during transplantation due

to ischemia-reperfusion injury to the graft, leading to immunologic scarring, a persistent

and pathologic alteration to this cellular network [16, 17]. Understanding how

immunosuppressants affect these architectural elements of immunity is crucial for

enhancing long-term allograft survival.

Apart from the effects on specific immune cell populations, immunosuppressive drugs

significantly impact the gut microbiome composition and their functional output,

influencing alloimmunity and graft survival [18-22]. Our group has demonstrated that gut

microbiota modulates LN architecture through alterations in stromal fiber laminins [23-

25], establishing a crucial link between the gut microbiome and LN structure and

function. Further, we observed significant alterations in both gut microbiome

composition and metabolic activities under tacrolimus-induced immunosuppression in

both luminal and systemic measurements [26]. Our recent work on inhibiting the

mammalian target of rapamycin (mTOR) revealed its broad effects on gut microbiome

structure, metabolic functions, and intestinal host responses, including changes in LN

architecture [27][28]. Other studies have shown diverse mechanisms by which

immunosuppressants affect host-microbe interactions. For instance, methotrexate

inhibits the conserved dihydrofolate reductase pathway in both host and gut bacteria,

leading to broad microbiome alterations [29], while other immunosuppressants modify

the intestinal environment through changes in barrier permeability [30]. Importantly,

these host-microbe interactions are multidirectional, in that microbiota can alter

intestinal function and drug bioavailability [31, 32]. For example, mycophenolate mofetil

(MMF) is metabolized to inactive glucuronidated mycophenolic acid (MPAG) via hepatic

glucuronidation, and luminal bacteria expressing β-glucuronidase can reactivate the

metabolized form MPAG back to active MPA, leading to increased concentrations of

active luminal drug and colonic inflammation [33]. Given these complex interactions

between immunosuppressants, gut microbiota, intestinal function, and immune cell

populations, understanding their intricate relationships is crucial to decipher both

therapeutic effects and unintended drug effects.

In this study, we systematically investigated the previously understudied effects of

immunosuppressants on LN architecture, the distribution of lymphocyte and myeloid cell

populations, and gut microbiome and metabolome. Four key classes of clinical

immunosuppressants were included: TAC (tacrolimus) from calcineurin inhibitor

category, PRED (prednisone) from glucocorticoid category, MMF from inosine

monophosphate dehydrogenase (IMPDH) inhibitor category, and FTY720 (Fingolimod)

from S1P receptor antagonist category. Dynamic temporal drug effects were observed,

where moderate early changes progressed to substantial alterations in the gut

microbiome and metabolism after prolonged treatment. By day 30, despite their distinct

mechanisms of action, all drugs converged to induce a similar dysbiotic state in the gut

microbiome. This convergence, marked by enrichment of pathobionts Muribaculaceae,

suggests a common mechanism linking long-term immunosuppression to metabolic

complications in transplant patients. Further, early treatment induced pro-inflammatory

changes in LN architecture and cellular composition, while long-term treatment resulted

in anti-inflammatory effects. Notably, in alloantigen-stimulated conditions, MMF and

FTY720 suppressed alloantigen-induced pro-inflammatory changes through dual

regulation of Tregs and LN architecture. These findings demonstrate complex dynamics

among immunosuppressants, microbiome changes, and immune tissue organization,

providing new insight into mechanism of action of long-term immunosuppressant use.

Results

Immunosuppressant treatments affect gut microbiome composition and function

We treated groups of C57BL/6 mice daily with four immunosuppressants (TAC, PRED, MMF and FTY), respectively, using untreated mice as controls. We analyzed the gut microbiome at early (3 days), intermediate (7 days), and late (30 days) time points. We performed whole community metagenomic sequencing on intraluminal stool samples collected from the colon and utilized the comprehensive mouse gut metagenome catalogue (CMGM) [34] (**Supplemental Table 1A**) for taxonomic profiling and Human Microbiome Project Unified Metabolic Analysis Network (HUMAnN2) (v0.11.2) [35] (**Supplemental Table 1B**) to analyze microbial metabolic pathways.

To evaluate drug effect on the gut microbiome, we calculated Bray-Curtis pairwise distances for taxonomic composition, comparing drug-to-control deviations from the untreated state and drug-to-drug differences between treatments. In drug-to-control comparisons, the effects were variable during early treatment by day 3, with some drugs increasing and others decreasing relative to controls (**Figure 1A**). However, by day 30, all treatment groups showed profound and significant alterations in their gut microbiome profiles, suggesting a cumulative effect of sustained immune suppression. The pattern was evident in both taxonomic composition and functional pathway analyses, with the most substantial changes emerging after prolonged treatment (**Figure 1B**). Drug-to-drug comparison revealed a pattern of temporal progression. The differences between treatment groups initially increased from day 3 to day 7, suggesting distinct early drug-specific effects on gut microbial communities (**Figure 1A**). FTY demonstrated milder

effects compared to other drugs during this period. By day 30, these inter-drug

differences were substantially reduced, indicating a convergence of the effect of long-

term immunosuppressant drugs on gut microbiome composition. This convergence was

observed in both taxonomic profiles and functional pathway abundances (**Figure 1B**),

suggesting that despite their different mechanisms of action, these distinct drugs

ultimately drive the gut microbiome toward a similar altered state. These findings were

further validated through β-diversity analyses. Principal Component Analysis (PCA)

revealed clear clustering of 30-day samples, distinct from earlier time points, regardless

of the specific immunosuppressant used (**Figure 1C**). Canonical Correspondence

Analysis (CCA) confirmed these observations, demonstrating significant correlations

between treatment duration and microbiome composition, with maximal separation

occurring at 30 days while day 3 and 7 samples showed minimal divergence from

controls (**Supplemental Figures 1A and 1B**). The temporal convergence of

microbiome alterations across different immunosuppressant drugs revealed shared

downstream influences of these drugs on host-microbe interactions, despite their

diverse initial effects.

To identify specific microbial signatures associated with immunosuppression, we

identified differentially abundant bacterial species between treatment groups and

controls at each time point (**Figure 1D**). No individual taxonomic group showed

consistent alterations across any time point in any treatment group. During early (day 3)

and intermediate (day 7) time points, we observed sporadic alterations in various

bacterial taxa, but these changes were transient and not maintained through the late

time point (day 30). By day 30, all treatment groups, despite their distinct mechanisms

of action, converged on a common set of bacterial taxa that were not significant at the

earlier time points. These taxa belonged exclusively to the family of Muribaculaceae

(formerly known as S24-7), a predominant commensal group in the mouse gut

microbiome, including *Duncaniella*, *Paramuribaculum intestinale*, and CAG-873,

pathobionts that are both necessary and sufficient to induce diabetes [36]. Together, the

data reveal time-dependent pattern of immunosuppressant effects. The convergent

enrichment of pathobionts suggests that prolonged immunosuppression promotes a

dysbiotic state that may contribute to the metabolic complications commonly observed

in transplant patients.

<u>Immunosuppressants induce drug-specific alterations in gut metabolic profiles</u>

We next examined gut luminal metabolism at day 7, a timepoint our previous work

identified as capturing rapid metabolic changes that precede microbiome alterations [26],

indicating direct effects on both host and microbial metabolism. Using untargeted

metabolomic assessment through capillary electrophoresis-mass spectrometry (CE/MS),

we identified 264 distinct metabolites in the gut lumen after quality assessment

(**Supplemental Table 2A-B**). Partial Least Squares Discriminant Analysis (PLS-DA)

demonstrated clear separations of metabolic profiles among the different

immunosuppressant treatments (**Supplemental Figure 2A**) with distinct top contributing

metabolites for each group (**Supplemental Figure 2B)**. Hierarchical clustering

heatmaps also showed distinct drug induced patterns of differentially abundant

compounds (**Supplemental Figure 2C**), indicating the distinct drug-specific metabolic

profiles in gut lumen after immune suppressant treatment.

Pairwise comparison of individual drugs to the no-treatment control, combined with

enrichment analyses, revealed both shared and drug-specific effects on metabolic

activities (**Table 1**, **Supplemental Table 2C-F**). TAC exhibited the most balanced

metabolic impact, showing both significant increases and decreases across multiple

pathways (**Figure 2A, Supplemental Table 2F**). TAC treatment led to marked

increases in three major categories of metabolites: 1) elevated levels of multiple amino

acids and their derivatives, indicating alterations in protein catabolism and/or amino acid

utilization in both host and microbiome. 2) TAC enhanced pyrimidine metabolism

suggesting increased nucleotide turnover. 3) Elevated sugars and sugar phosphates

indicate enhanced glycolytic pathway activity and altered central carbon metabolism.

TAC treatment also significantly decreased several key metabolic pathways. These

included reduced levels of polyamines and their derivatives, biogenic amines, short-

chain fatty acids (SCFAs), betaine, trimethylamine (TMA), and various organic acids.

PRED, MMF, and FTY demonstrated predominantly suppressive impacts on gut

metabolism, with distinct patterns and scope. PRED exerted the most broad and

profound suppressive effect, decreasing three to four times more compounds than other

treatments while showing minimal metabolite increases (**Figure 2B, Supplemental**

Table 2D). A broad spectrum of metabolites across multiple pathways were reduced,

including amino acid derivatives, SCFAs, other microbial-derived metabolites like TMA,

pyrimidine components, and polyamines. This suppressive pattern aligns with known

widespread effects of PRED on cellular metabolism.

MMF showed a predominant suppressive effect but with more selective targeting of

specific pathways (**Figure 2C, Supplemental Table 2E**). Key suppressed pathways

include glyoxylate and dicarboxylate metabolism involved in the TCA cycle, branched-

chain amino acids (BCAAs), tyrosine metabolism, and purine metabolism. MMF

increased adenosine while reducing uric acid, consistent with its mechanism as an

IMPDH inhibitor. This may reflect increased IMP availability leading to enhanced flux

through adenosine pathways, directly linking metabolic changes to the MMF therapeutic

mechanism. On the other hand, FTY demonstrated a moderate suppressive effect

Figure 2D, Supplemental Table 2C). FTY increased nucleoside levels while

decreasing nucleotides and SCFAs. It reduced arginine availability while increasing

N^1,N^8-diacetylspermidine and spermidine, indicating a directed metabolic processed. In

summary, these results demonstrate distinct drug-specific metabolic signatures in the

gut environment: TAC's balanced modulatory effects across multiple pathways, PRED's

broad and profound suppression, MMF's selective suppressive effect, and FTY's

moderate changes focused on nucleotide metabolism. These data indicate that the

significant metabolic alterations may contribute to both therapeutic effects and side

effects of different immunosuppressive strategies.

<u>Immunosuppressants elicit time-dependent, drug-specific intestinal responses</u>

To investigate the reciprocal interactions between the altered gut microbiome and gut

epithelium responses under drug treatment, we analyzed the intestinal transcriptome.

Days 7 and 30 were chosen as they corresponded to the major gut microbiota

alterations during drug treatments. Differentially expressed genes (DEGs) were

identified by comparing each drug group to the no-treatment control (**Supplemental**

Table 3A). Analysis of DEGs showed that 30-day treatment altered 1.3 to 2 times more

genes than 7-day treatment, with the majority showing decreased expression levels.

This pattern suggests prolonged immunosuppression leads to more extensive

transcriptional modulation in intestinal tissue. The magnitude of response elicited by

each individual drug was distinct (**Figure 3A, Supplemental Table 3B**). At day 7, FTY

resulted in the largest number of DEGs. By day 30, both FTY and PRED elicited

stronger effects in terms of upregulation and downregulation of DEGs. TAC showed the

most unbalanced response, with twice as many genes showing increased expression

compared to decreased expression. In contrast, PRED and MMF treatments produced

more balanced alterations in gene expression patterns, with a greater proportion of

genes showing decreased expression levels at day 30 compared to day 7. This

temporal shift indicates that prolonged treatment leads to broader changes in intestinal

gene expression.

We used UpSet plots to analyze overlapping DEGs between days 7 and 30 and across

different treatments. These plots revealed minimal shared genes between timepoints

and across different immunosuppressants (**Supplemental Figure 3A-D, Supplemental**

Table 3C-F). For instance, following PRED treatment, only 31 genes (15 upregulated,

16 downregulated) are consistently modulated at both timepoints; there are 255 genes

uniquely regulated at day 7 (81 upregulated, 78 downregulated) and 371 genes

uniquely modulated at day 30 (174 upregulated, 197 downregulated). This pattern holds

true for all four drug treatments. The limited overlap in DEGs between the two

timepoints suggests these immunosuppressants induced distinct responses at day 7

and 30. Further, the drug impact is more pronounced at day 30 with greater number of

DEGs at either up- or down-regulation, indicating that long-term immunosuppression

involves increasingly distinct cellular processes for each drug. Analysis of DEG between

treatment groups revealed that while these immunosuppressants shared some common

regulatory targets, they predominantly acted through distinct patterns of gene

expression (**Supplemental Figure 4A-B, Supplemental Table 3F-G**). The limited

overlap in regulated genes, typically less than 20% between any two drugs and less

than 5% across all four drugs, suggests highly drug-specific mechanisms of action. The

findings further support that despite their common therapeutic goals, these

immunosuppressants achieve immune modulation through largely independent

molecular regulation in intestine.

To characterize intestinal immune responses, we combined RNA-seq pathway analysis

with immunohistochemical (IHC) assessment of key immune populations. This

approach allowed us to examine both broad transcriptional changes and specific

alterations in CD11c+ dendritic cells (DCs) and Foxp3+ regulatory T cells (Tregs), two

cell types critical for immune homeostasis. GO ontology analyses of immune-related

pathways in DEGs revealed distinct temporal patterns among the four drugs (**Figure**

3B). At day 7, FTY decreased expression of genes involved in mucosal and humoral

immunity, while other drugs showed moderate effects on B cell-mediated response

pathways. By day 30, the pattern shifted dramatically: FTY increased expression across

multiple immune pathways; MMF switched from early activation to broad reduction in

innate immune pathway expression; TAC progressed from moderate increases in

lymphocyte and immunoglobulin pathways to stronger activation of both innate immunity

and antibody production; and PRED maintained relatively modest effects on immune

pathway expression throughout the treatment period. IHC analysis revealed that the

drug effects were similarly time-dependent but transient (**Supplemental Figure 5A-C)**.

For intestinal CD11c+ DCs, all four drugs decrease these at day 7 but had no effects on

days 3 or 30. For intestinal Foxp3+ Tregs, there were some increases on day 3 (by

TAC and FTY), some decreases on day 7 (by TAC, PRED, and FTY), and no effects by

day 30. Together, these findings demonstrate that immunosuppressants induce

complex time-dependent changes in intestinal immune responses, with each drug

showing distinct patterns of gene expression and immune cell modulation that evolve

over the course of treatment.

<u>Time-dependent impact of immunosuppressants on mesenteric and peripheral lymph</u>

<u>nodes (mLN and pLN) organization</u>

We next investigated how immunosuppressants affected LN organization in both

mucosal and systemic compartments by examining mLN and pLN, the primary draining

sites for mucosal and systemic immunity, respectively. Using flow cytometry and IHC,

we analyzed changes in the numbers and distribution of immune cell populations

(B220+ B cells, CD4+ T cells, CD8+ T cells, Foxp3+ Tregs) and structural features

La4:La5 in LNs. We particularly assessed the distribution of La4:La5 and Tregs in the

CR and around the HEV of LNs to understand how immunosuppressants modulated

cellular and structural aspects of immune organization.

In mLN, flow cytometry revealed that B cells, CD4+ T cells, CD8+ T cells and Foxp3+

Tregs were unaffected by TAC or MMF treatment at days 3, 7, or 30 (**Figure 4C**). In

contrast, PRED treatment significantly reduced CD4+ T cells and B220+ B cells by day

30, with no significant changes at earlier time points (**Figure 4A, C**). FTY treatment

increased Foxp3+ Tregs by day 30, with no changes at days 3 or 7 (**Figure 4B, C**).

Overall, while CD8+ T cell populations remained not significantly changed across all

conditions (data not shown), other populations showed treatment-specific changes that

were only pronounced after prolonged use. IHC results revealed two patterns of

changes. On one hand, after TAC or PRED treatment, no significant changes in Treg

distribution were observed in the mLNs at any time points (**Figure 4D, E, H, I,**

Supplemental Figure 6A, B). However, both drugs reduced the La4:La5 around the

CR after 7 days (**Figure 4F, H, I, Supplemental Figure 6C**) and TAC also reduce it in

the HEV after 7 days (**Figure 4G-I, Supplemental Figure 6D**). On the other hand, MMF

and FTY showed similar patterns with an increased Treg distribution in the CR by day 7

followed by a decrease by day 30 (**Figure 4D, H, I, Supplemental Figure 6A**). They

also decreased La4:La5 in both the CR and HEV (**Figure 4F-I, Supplemental Figure**

6C, D). The responses at day 3 were mixed, indicating unstable early reactions that

converged in later time points after prolonged drug treatment. Overall these data

indicate that despite their distinct mechanisms of action, all four immunosuppressants

converge on promoting a more pro-inflammatory environment in mLN by day 30,

characterized by reduced Treg presence and/or decreased La4:La5. Transitional

changes began around day 7, with early changes at day 3 being variable. This dynamic

pattern suggests that prolonged immunosuppression may paradoxically create localized

pro-inflammatory conditions in mucosal lymphoid tissues.

In pLN, distinct temporal patterns of immune cell modulation were observed. TAC and

PRED had minimal effects on cellular composition (**Figure 5D**), while MMF selectively

reduced CD4+ T cells by day 30 (**Figure 5A, D**). FTY showed the most dynamic

changes: increasing B220+ B cells while decreasing CD4+ T cells at day 7 (**Figure 5B**),

followed by elevated Foxp3+ Tregs and decreased CD4+ T cells by day 30 (**Figure 5C,**

D). IHC analysis revealed drug-specific effects on both cellular distribution and

structural features. TAC showed transient effects, increasing Treg distribution in the CR

at day 7 (**Figure 5E, I, Supplemental Figure 7A, B**) without affecting laminin ratios

(**Figure 5G-J, Supplemental Figure 7C, D**). PRED demonstrated only early structural

changes by day 3 and only increases La4:La5 **(Figure 5E-J, Supplemental Figure 7A-**

D). MMF showed the most consistent anti-inflammatory profile, maintaining increased

Treg distribution **(Figure 5E, F, I, J, Supplemental Figure 7A, B)** throughout the

treatment period and increasing La4:La5 at day 3 **(Figure 5G-J, Supplemental Figure**

7C, D). FTY showed only early anti-inflammatory effects, increasing both Treg

distribution and La4:La5 at day 3 but not the later time points **(Figure 5E-J,**

Supplemental Figure 7A-D). Overall, these data reveal that immunosuppressants

predominantly promote an anti-inflammatory state in pLN, characterized by increased

Tregs and/or elevated La4:La5, particularly during early treatment phases.

MMF and FTY720 counteract alloantigen-induced pro-inflammatory changes through

dual regulation of Tregs and LN architecture

Given the significant impact noted above of MMF and FTY on LN architecture with

increasing Tregs and La4:La5 under homeostatic conditions (**Figure 4H, I, 5I, J**), we

investigated their effects in an inflammatory context using allogeneic stimulation. Mice

received allogeneic splenocytes (10^7 cells intravenously) with or without drug treatment,

and responses were assessed after three days. While allogeneic cells alone did not

affect Treg distribution in mLN or pLN (**Figure 6A, C**), the combination of drugs with

allogeneic cells showed significant effects: MMF increased Tregs around mLN HEV,

while FTY increased Treg in both mLN and pLN HEV and CR (**Figure 6A, C, E, F**).

Allogeneic stimulation alone decreased La4:La5 in both pLN and mLN (**Figure 6B, D**),

indicating a pro-inflammatory LN structural shift in the context of alloantigen-induced

immune responses, consistent with our previous findings [15]. Both MMF and FTY

treatments prevented this decrease, maintaining La4:La5 comparable to untreated

controls **(Figure 6B, D, E, F)**. These findings demonstrate that MMF and FTY inhibit

alloantigen-induced proinflammatory changes in LN organization through two

complementary mechanisms: increasing Tregs and preserving tolerogenic laminin

stromal architecture. This dual action on both T cell and stromal cell structural features

suggests these drugs may be particularly effective at promoting a tolerogenic

environment during allostimulation.

MMF and FTY effects are mediated through FRC derived laminins α4 and α5

Given the significant impact of MMF and FTY on LN architecture by affecting La4:La5

and the crucial role of FRC in laminin expression [14, 37], we investigated whether the

effects of these drugs on LNs are mediated through FRC-derived laminins. We utilized

two laminin KO mouse strains: FRC-specific laminin α4 knockout (FRC-Lama4-KO;

Pdgfrb-Cre+/- × Lama4$^{fl/fl}$) and FRC-specific laminin α5 knockout (FRC-Lama5-KO;

Pdgfrb-Cre+/- × Lama5$^{fl/fl}$) mice. These models allowed us to specifically examine the

role of FRC-derived laminin α4 and α5 in immunosuppressant-mediated effects on LN

stromal organization.

In mLN, as noted above, 3-day treatment of WT mice with MMF or FTY did not affect

laminin ratios in the mLNs (**Figure 4F-I, Supplemental Figure 6C, D**). MMF and FTY

also did not affect laminin ratios in FRC-Lama4-KO mice (**Figure 7A-C, G**). However, in

FRC-Lama5-KO mice, both drugs decreased the La4:La5 in CR of mLNs (**Figure 8D-G**).

These results suggested that MMF and FTY effects on laminin composition is

dependent on laminin α5 expression, indicating its critical role in maintaining the laminin

balance in the mLN.

In pLNs, as noted above, 3-day treatment of WT mice with MMF or FTY each increased

La4:La5 in the pLNs (**Figure 5G-J, Supplemental Figure 7C, D**). MMF also increased

La4:La5 in both CR and HEV of FRC-Lama5-KO mice (**Figure 8D-G**), but not in FRC-

Lama4-KO mice (**Figure 8A-C, G**). In contrast, FTY decreased La4:La5 in CR of FRC-

Lama4-KO mice (**Figure 8A-C, G**) but not in FRC-Lama5-KO mice (**Figure 8D-G**).

These findings indicated that MMF and FTY modulated La4:La5 through distinct

mechanisms that depended on the presence of laminins α4 and α5 from FRCs,

although the precise pathways involved remain to be determined.

Discussion

This study provides an in-depth characterization of the previously underappreciated

complex interplay of immunosuppressants with the gut microbiome and metabolism,

intestinal gene expression, mesenteric and peripheral LN organization, and lymphocyte

trafficking over time. We observed prolonged immunosuppressant treatment leads to a

convergent dysbiotic state in the gut microbiome by day 30, despite distinct drug-

specific effects observed at earlier time points. This incremental drug effect suggests a

common underlying mechanism related to immune suppression, rather than individual

drug effects, that impacts gut microbial communities. In the context of the metabolic and

intestinal response results, the influence on gut microbiome likely related to nutrient

availability, immune modulation, and altered gut function, which favors a particular

microbial community structure. Particularly notable is the consistent enrichment of

Muribaculaceae family members by day 30, known mediators of metabolic disorders

such as diabetes [36], suggesting a potential mechanistic link between long-term

immunosuppression and metabolic complications commonly observed in transplant

patients [38-40].

Each immunosuppressant exhibited distinct metabolic signatures that revealed both

expected and unexpected mechanisms of action. TAC broadly alters amino acid

metabolism and other microbial-derived metabolites such as SCFAs, suggesting an

influence on gut microbiome. Previous studies reported gastrointestinal side effects like

diarrhea and nonspecific abdominal discomfort after tacrolimus use [41]. PRED exerted

a broad suppressive effect on metabolism, consistent with its known ability to inhibit

cellular functions and pro-inflammatory cytokines [42]. This reduction in amino acid

derivatives, SCFAs, and polyamines provides insights into its metabolic side effects,

including diabetes and osteoporosis [43-45]. The reduction in microbial-derived

metabolites such, as TMA and SCFAs, suggests substantial disruption of host-microbe

interactions. MMF selectively targeted specific metabolic pathways, revealing a more

precise mechanism compared to PRED and TAC. Its effects on purine metabolism

directly align with its known IMPDH inhibition. The impact on the TCA cycle and BCAA

metabolism might explain its distinct side effect profile of gastrointestinal toxicity and

inflammation [46, 47]. Previous studies report that by inhibiting the TCA cycle, MMF

reduces ATP production in intestinal epithelial cells, impairing barrier integrity through

disrupted tight junction protein assembly [30]. Its disruption of BCAA metabolism

compromises enterocyte protein synthesis and repair, leading to increased enterocyte

apoptosis and villous atrophy [48]. FTY showed the most moderate metabolic effects,

focused primarily on nucleotide metabolism and specific amino acid pathways. The

moderate suppressive effect with maintenance of other metabolic pathways might

explain its generally better tolerability. These distinct metabolic signatures among drugs

could influence nutrient availability, signaling molecules, and microbial substrates,

thereby shaping separate ecological niches within the gut[49].

Notably, all drugs affected SCFA levels to some degree, suggesting a common impact

on microbiota-dependent metabolic functions. This shared effect may contribute to the

convergent microbiome changes observed by day 30 and could represent a

fundamental mechanism underlying long-term immunosuppression. Reduction in SCFA-

producing bacteria can disrupt metabolic processes, influencing blood pressure

regulation, lipogenesis, and blood glucose control [50]. It was shown that probiotic

administration could reverse gut dysbiosis, particularly catabolic microbial profiles that

involve sucrose and starch degradation [50]. Other studies have shown that probiotics

can improve tacrolimus-induced gut dysbiosis, SCFA depletion, and hypertension by

restoring the vascular redox state and improving endothelial nitric oxide synthase

(eNOS) function [51, 52]. Differences in how these drugs alter amino acid metabolism

suggest they exert distinct influences on protein synthesis and cellular energy balance,

potentially explaining their varying effects on different immune cell populations. Arginine

metabolism is particularly relevant, as it affects eNOS activity through the arginine-NO

pathway, influencing NO production and reactive oxygen species (ROS) generation.

These changes can impair endothelial function and modulate immune cell activation.

We show that TAC and mTOR inhibitors like rapamycin drastically affect amino acid

regulation [REF], including arginine metabolism, and both have documented vascular

complications [53-55]. Research in restoring drug induced metabolic and microbial

dysregulation is still at very early stages, and our findings emphasize the importance to

mitigate the unintended effects on the gut microbiome and metabolism. They also

support the value of metabolic profiling and longitudinal monitoring to inform

personalized treatment strategies.

The gut serves as a critical interface connecting immunity to environmental stimuli,

making its regulation essential for maintaining transplant tolerance [56-58]. Our findings

indicate that immunosuppressants exert complex, time-dependent effects on the gut

environment, influencing immune responses, metabolic activities, and microbial

communities. While the gut microbiome converged toward a common dysbiotic state by

day 30, intestinal gene expression patterns became increasingly drug-specific, with

minimal overlap in differentially expressed genes (<5% across drugs). The progressive

divergence in gene expression profiles by day 30 indicated increasingly specialized

cellular processes unique to each drug. This result means that even as microbial

communities arrived at a similar configuration, the gut tissue itself continued to respond

in uniquely tailored ways to each immunosuppressant. Such tissue-specific

transcriptomic divergence likely reflects the direct drug actions on host cells, the host's

compensatory adjustments, and bidirectional interactions with microbiome metabolism.

Understanding these complex interplays emphasizes the need for monitoring and

potentially mitigating these effects in clinical practice.

Our study reveals the anatomically distinct and temporally dynamic effects of

immunosuppressants on lymphoid tissue organization. One key observation is the

divergent effect between mucosal and peripheral LN compartments. The pLN

maintained an anti-inflammatory state, characterized as elevated Tregs and La4:La5,

especially during early treatment phases. The pLN response aligns with the intended

immunosuppressive effects of these drugs. On the other hand, in mLNs, which drain the

intestine, all four drugs ultimately promoted a pro-inflammatory environment,

characterized by reduced Tregs and decreased La4:La5. This compartment-specific

regulatory effect may suggest a potential mechanism for some gastrointestinal side

effects seen with immunosuppression [26].

The crucial role of FRC-derived laminins α4 and α5 in mediating these effects indicated

a stromal response mechanism in immunosuppression driven laminin α4 and laminin α5

responses [14]. MMF and FTY both preserved tolerogenic LN architecture during

allostimulation through the dual regulation of Tregs and stromal organization, while they

were distinctively dependent on laminin α4 or laminin α5. Understanding these

anatomically distinct and mechanistically diverse effects could lead to more precise

immunosuppressive strategies that better maintain tolerogenic niches. Specifically, the

precise effects of different immunosuppressive drugs on the gene expression, function

and phenotype of LNSCs is a major gap in our knowledge and requires further

investigations.

The major limitations in this study are our focus on mouse models and relatively short

duration of study. The convergent effects at day 30 might not fully represent human

responses, and longer-term studies may better reflect clinical scenarios. Additionally,

we examined single drugs rather than combinations commonly used in practice. Our

findings suggest specific therapeutic strategies such as metabolic monitoring to predict

drug responses, targeted interventions to preserve microbiome dysbiosis, and tissue-

specific drug delivery to optimize local effects while minimizing systemic complications.

Methods and Materials

Study approval

All procedures involving mice were performed in accordance with the guidelines and regulations set by the Office of Animal Welfare Assurance of the University of Maryland School of Medicine under the approved IACUC protocol nos. 05150001, 0318002, 1220001, and AUP-00000397.

Reagents

USP grade immunosuppressants were used if available. All drugs were prepared in a manner approved by the University of Maryland School of Medicine IACUC: TAC (USP grade, MilliporeSigma, Burlington, MA) was reconstituted in DMSO (USP grade, MilliporeSigma) at 20 mg/ml for storage. At time of use, TAC/DMSO stock was diluted with absolute ethanol (USP grade, Decon Labs, King of Prussia, PA) to 1.5 mg/ml then further diluted 1:5 in sterile PBS and injected at 10 µl/gm s.c. (3 mg/kg/day)[25]. PRED (MilliporeSigma) was reconstituted in PBS at 50 mg/ml for storage and then diluted to 0.5 mg/ml in PBS and injected at 5mg/kg/d i.p.[59]. MMF (MilliporeSigma) was reconstituted at 30 mg/ml in DMSO for storage and then diluted in PBS to 1.5 mg/ml and at 30 mg/kg/d i.p.[60]. FTY (Fingolimod, MilliporeSigma) was reconstituted at 3 mg/ml in 1:1 PBS:Ethanol for storage and then diluted 1:10 in PBS and administered at 3 mg/kg/d p.o.[61].

Mice

8- to 14-week-old female C57BL/6 mice between were purchased from The Jackson Laboratory (Bar Harbor, ME, USA). Mice were maintained at the University of Maryland

School of Medicine Veterinary Resources animal facility. Only female mice were used

to ensure a high degree of homogeneity within our study groups. All procedures

involving mice were performed in accordance with the guidelines and regulations set by

the Office of Animal Welfare Assurance of the University of Maryland School of

Medicine. Mice received daily immunosuppression following the dosages in the table

above. During arrival at the animal facility, all mice were cohoused and handled

together to ensure all groups were exposed to each other but separated during

immunosuppressant treatment. CO_2 narcosis was utilized for euthanasia on the day of

harvest. We collected intralumenal stool samples for metagenomic and metabolomic

analyses. Blood collection was carried out using cardiac puncture at the time of

euthanasia We acknowledge the potential effects of first-pass hepatic metabolism on

general circulation, however, over extended periods, systemic metabolites should

normalize, unless there are acute changes in the intestine that we did not observe

within the experimental timeframe. Portal blood provides a more direct measure of

metabolites absorbed from the intestine before they undergo hepatic metabolism.

However, in mice, portal blood can only be reliably obtained at the time of euthanasia,

and the obtainable volumes are exceedingly low, making its routine collection

challenging. We collected mLN, pLN, and portions of the small intestine between the

duodenum and jejunum, for immunohistochemistry. The mLN and pLN samples were

harvested for flow cytometry analyses. Mouse experiments were performed according

to ARRIVE guidelines (https://arriveguidelines.org).

<u>Stool specimen collection, DNA extraction, and metagenomic sequencing</u>

Mice jejunum and colon tissues were dissected according to their gastrointestinal

anatomical features. From the dissected tissues, intraluminal stool contents were

collected and stored immediately in DNA/RNA Shields (Zymo Research, Irvine, CA,

USA) at -80°C. This stabilizes and protects the integrity of nucleic acids and minimizes

the need for immediate processing or freezing of specimens. DNA extraction was

described previously [62, 63]. In brief, the Quick-DNA Fecal/Soil Microbe kit (Zymo

Research, Irvine, CA, USA) was used to extract 0.15-0.25 grams of fecal samples. To

ensure that no exogenous DNA contaminated the samples, we includednegative

extraction. Construction of metagenomic sequencing libraries was performed using the

Nextera XT Flex Kit (Illumina), according to the manufacturer's recommendations. Then,

libraries were pooled together in equimolar proportions before they were sequenced on

a single Illumina NovaSeq 6000 S2 flow cell at Maryland Genomics at the University of

Maryland School of Medicine.

<u>Gut microbiome analyses</u>

We removed metagenomic sequence reads using BMTagger v3.101 [64] mapping to

Genome Reference Consortium Mouse Build 39 of strain C57BL/6J (GRCm39) [65].

When one or both of the read pairs matched the genome reference, sequence read

pairs were removed. Trimming and quality assessment of the Illumina adapter was

performed using default parameters in fastp (v.0.21.0) [66]. We established taxonomic

composition of the microbiomes using Kraken2 (v.2020.12) [67] and Braken (v. 2.5.0)

[68] and the comprehensive mouse gut metagenome catalog (CMGM) [34] to calculate

the metagenomic taxonomic composition. Barplots and diversity indices were generated

using Phyloseq R package (v1.38.0) [69]. We employed the Chao1 diversity index, an

abundance-based indicator of species richness (total number of species in a sample)

that is sensitive to low abundance taxonomic groups (singletons and doubletons) [70]

because the mice gut microbiome datasets contain a broad spectrum of taxa without

any dominant high-abundance species. This choice ensures accurate representation

and understanding of the microbial community diversity in our samples. To identify fecal

phylotypes that could explain the differences, we utilized linear discriminant analysis

(LDA) effect size (LEfSe) analysis [71]. For the non-paramentric factorial Kruskal Wallis

(KW) sum-rank test, the α value was set at 0.05 and the threshold for the logarithmic

LDA model [72] score for discriminative features was set at 2.0. In the multiclass

analysis, an all-against-all BLAST search was performed. The limma voom function [73]

in R package microbiomeMarker v1.3.3 [74] was used to calculate microbial biomarkers.

Phylogram representing the taxonomic hierarchical structure of the identified phylotype

biomarkers via pairwise comparisons between groups, graph generated using R

package yingtools2 [75]. To ensure comprehensive coverage in functional annotation,

the metagenomic dataset was mapped to the protein database UniRef90 [76]. To

efficiently and accurately determine the presence, absence, and abundance of

metabolic pathways in a microbial community, the dataset was then summarized using

HUMAnN3 (Human Microbiome Project Unified Metabolic Analysis Network)

(v3.0.0.alpha.3) [35]. Canonical Correspondence Analysis (CCA) utilizing the vegan

package [77, 78] based on the Bray-Curtis distance was used for ordination analysis.

Based on their eigenvalues, CA1 and CA2 were selected as the major components. To

evaluate microbiome profile differences across groups and time points, the line plot was

generated using the average pairwise Bray-Curtis distances between all group pairs at

each time point.

<u>Metabolite extraction and metabolome analyses</u>

Capillary electrophoresis-mass spectrometry (CE/MS) was used for measuring

metabolome of intraluminal stool (luminal/local) to obtain a comprehensive quantitative

survey of metabolites (Human Metabolome Technologies, Boston, MA, USA) for

immunosuppressants TAC, PRED, MMF, and a no-treatment control group processed

concurrently. For FTY720, stool pellets were collected at the same time points for

metabolomic analysis due to technical limitations in obtaining sufficient sample volumes

for comprehensive profiling. The no-treatment control for FTY720 also used stool pellets

and processed concurrently. All subsequent analyses were performed using the

respective no-treatment control groups and the same specimen types, processed

concurrently with their corresponding experimental groups, to ensure valid comparisons.

~10-30 mg of stool was weighed at the time of collection using a company-provided vial

and stored at -80°C until shipped to vendor on dry ice. Strict QC includes blanks and

internal standards evenly spaced among samples and in every batch.

To reduce the influence of measurement noise, rigorous data pretreatment was

performed according to validated procedures [79]. Log base 10 transformation was

applied to individual values themselves, and data scaling adjusted each variable/feature

by a scaling factor computed based on the range of each variable as the dispersion of

the variable. Known metabolic databases or the *a priori* knowledge-based approach

were used to exhaustively annotate metabolites, achieved using PubChem [80], KEGG

(vers Dec 2023) [81], HMDB [82], and MetaboAnalyst 5.0 [83, 84] annotation

frameworks that leverage cataloged chemical compounds, known metabolic

characterization, and functional hierarchy (i.e., reaction, modules, pathways). Partial

least squares discriminant analysis (PLS-DA) implemented using mixOmics (vers.

6.18.1) was used [85]. To show significantly increased metabolites after 7-day

tacrolimus treatment, the volcano plot combines results from fold change (FC) analysis.

A metabolite is shown when the FC is >2 and the p-value is <0.05 based on 2-sample t-

tests. In the FC analysis, original metabolite measurements without normalization were

utilized. For the metabolites annotated in a specific functional pathway, metabolite set

enrichment analysis (MSEA) was performed [86].

RNA isolation, transcriptome sequencing and analyses of the intestinal tissues

The colon tissues of the mice were dissected according to their anatomic features and

stored immediately in RNAlater solution (QIAGEN) in RNAse-free 1.7-ml tubes (Denville

Scientific, Holliston, MA) at -80°C to stabilize and protect the integrity of RNA. For each

sample, total RNA was extracted from ~1 cm of ileum. Prior to the extraction, 500 μl of

ice-cold RNase free PBS was added to the sample. To remove the RNAlater, the

mixture was centrifuged at 8,000x*g* for 10 min and the resulting pellet resuspended in

500 μl ice-cold RNase-free PBS with 10 μl of β-mercaptoethanol. A tissue suspension

was obtained by bead beating procedure using the FastPrep lysing matrix B protocol

(MP Biomedicals, Solon, OH) to homogenized tissues. RNA was extracted from the

resulting suspension using TRIzol Reagent (Invitrogen, Carlsbad, CA) following the

manufacturer's recommendations and followed by protein cleanup using Phasemaker

tubes (Invitrogen) and precipitation of total nucleic acids using isopropanol. RNA was

resuspended in DEPC-treated DNAase/RNAase-free water. Residual DNA was purged

from total RNA extract by treating once with TURBO DNase (Ambion, Austin, TX, Cat.

No. AM1907) according to the manufacturer's protocol. DNA removal was confirmed via

PCR assay using 16S rRNA primer 27☐F (5′-AGAGTTTGATCCTGGCTCAG-3′) and

534☐R (5′-CATTACCGCGGCTGCTGG-3′). The quality of extracted RNA was verified

using the Agilent 2100 Expert Bioanalyzer using the RNA 1000 Nano kit (Agilent

Technologies, Santa Clara, CA). Ribosomal RNA depletion and library construction

were performed using the RiboZero Plus kit and TruSeq stranded mRNA library

preparation kit (Illumina) according to the manufacturer's recommendations. Libraries

were then pooled together in equimolar proportions and sequenced on a single Illumina

NovaSeq 6000 S2 flow cell at the Genomic Resource Center (Institute for Genome

Sciences, University of Maryland School of Medicine) using the 150 bp paired-end

protocol. An average of 87- 130 million reads were obtained for each sample. The

quality of Fastq files was evaluated by using FastQC [87]. Reads were aligned to the

mouse genome (Mus_musculus.GRCm39) using HiSat (version HISAT2-2.0.4) [88] and

the number of reads that aligned to the coding regions was determined using HTSeq

[89]. Significant differential expression was assessed using DEseq2 [90] with an FDR

value ≤ 0.05 and FC>2. The over-representative analysis was done by importing

Differentially Expressed Genes (DEGs) against GO ontologies using the enrichGO

function of clusterProfile Bioconductor package [91]. Only the ontology terms with q

value <0.05 were used for plotting. UpSet plots were generated using UpSetR (vers.

1.4.0) to analyze overlapping DEGs between days 7 and 30 and across different

treatments.

<u>Flow cytometry</u>

To produce single-cell suspensions, LNs were disaggregated and passed through 70-

µm nylon mesh screens (Thermo Fisher Scientific, Waltham, MA). Antibodies against

surface molecules were incubated with cell suspensions for 30 min at 4°C

(**Supplemental Table 4**) then washed with FACS buffer [PBS with 0.5% w/v bovine

serum albumin] twice. Cells were then permeabilized, according to manufacturer's

protocol, with Foxp3/Transcription Factor Staining Buffer Set (eBioscience, San Diego,

CA), washed with FACS buffer, and subsequently stained with antibodies for

intracellular molecules at 4°C. Samples were analyzed with an LSR Fortessa Cell

Analyzer (BD Biosciences), and data were analyzed using FlowJo software version 10.6

(BD Biosciences).

<u>Immunohistochemistry</u>

We excised mLN, pLN and segments of the intestine then immediately submerged in

OCT compound (Sakura Finetek, Torrance, CA, USA). Cryosections (6-7 µm for LNs,

10 µm for intestine) were cut in duplicate or triplicate depending on size of block using a

Microm HM 550 cryostat (ThermoFisher Scientific). Slides were then fixed in cold 1:1

acetone:methanol for 5 min, air dried for 15 min, and stored at -80°C until use. At time

of staining, slides were thawed for 15 minutes at room temperature then. sections were

rehydrated in PBS and blocked with10% secondary antibody host serum PBST (PBS +

0.03% Triton-X-100 + 0.5% BSA). The slides were then stained for 1 hour with primary

antibodies at room temperature (diluted 1:50-1:200 in PBST) for 1 hour incubated with

secondary antibodies (diluted 1:100-1:400 in PBST) for 1 hour. After staining, slides

were fixed with 4% paraformaldehyde in PBS for 5 min, quenched with 1% glycerol in

PBS for 5 min, and mounted with Prolong Gold Antifade Mountant with or without DAPI

(Thermo Fisher Scientific). Images were acquired using a Nikon Accu-Scope EXC-500

(Nikon, Tokyo, Japan) and analyzed using Volocity software (PerkinElmer, Waltham,

MA). The antibodies used are listed in **Supplemental Table 4**. More than 3 mice/group,

1-2 mLN or pLN or 3 pieces of intestine, 2-3 sections/tissue sample, and 7-30

fields/tissue sample were analyzed. We calculated mean fluorescence intensity (MFI)

within demarcated high endothelial venules (HEV) and cortical ridge (CR) regions of the

mLN and pLN as well as of intestinal images. Qualitative heat maps were generated

(GraphPad prism) to express changes in IHC marker expression level relative to control

using -1 to represent "decreased" (blue in the heatmap), 0 to represent "unchanged"

(white in the heatmap), and 1 to represent "increased" (red in the heatmap). GraphPad

Prism 10.3.1 was used for analysis (San Diego, CA, USA) with statistical significance

defined as $P < 0.05$. Tukey's multiple comparison tests of one-way ANOVA were used

for comparisons of fluorescent.

Declarations

Ethics Approval and Consent to Participate

All experimental procedures involving mice were approved by and performed in accordance with the guidelines and regulations of the Institutional Animal Care and Use Committee (IACUC) of the University of Maryland School of Medicine under the approved protocol numbers 05150001, 0318002, 1220001, and AUP-0000397.

Consent for Publication

Not applicable.

Availability of data and materials

The authors confirm that the data supporting the findings of this study are available within the article and indicated supplementary materials. The data that support the findings of this study are openly available; metagenome sequences were submitted to GenBank under BioProject PRJNA809764 (https://www.ncbi.nlm.nih.gov/bioproject/PRJNA809764) with the SRA study ID SRP361281. The SRA accession numbers for each sample and their biosample ID are included in **Supplemental Table 1**. The R codes, including each step and parameters, were deposited in github at https://github.com/igsbma/omics_paper23. Supplemental data for this article can be accessed online at https://figshare.com/account/home#/projects/173364

Competing interests

No potential financial and non-financial conflict of interest was reported by the authors.

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(SJG).

Authors' contributions

B. M., and J. S. B. designed the experiments. L. W., A. K., S. J. G., V. S., W. P., R. L., J.

I., M. W., and L. L. conducted and analyzed the mice experiments. H. W. L. performed

nucleic acid extraction and sequencing library preparation. B. M. and Y. S. performed

the bioinformatics analyses. B. M., L. W., A. K., S. J. G., M. F., and J. S. B. wrote the

manuscript. B. M. and J. S. B. are co-corresponding authors for this manuscript.

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multidrug transporter- and cysteinyl leukotriene-dependent T cell chemotaxis to
lymph nodes. *Journal of Clinical Investigation* 2003, **111**(5):627-637.
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1012 **Tables**

1013 **Table 1**. Differential metabolite abundance in gut lumen after 7 days of

1014 immunosuppressant treatment *

	TAC	PRED	MMF	FTY720
↑	**31** • Amino acids & derivatives • Pyrimidine metabolism • Sugars and sugar phosphates	**4** • Amino acids & derivatives	**4** • Sugar phosphates • Adenosine	**14** • Purine metabolism • Arginine metabolism
↓	**28** • Polyamines & derivatives • SCFAs, TMA	**95** • Amino acids & derivatives • SCFAs, TMA • Pyrimidine metabolism • Polyamines & derivatives	**37** • Glyoxylate and Dicarboxylate • Branched-chain amino acids & derivatives • Purine metabolism • Neurotransmitters	**22** • Pyrimidine metabolism • SCFAs
subtotal	**59**	**99**	**41**	**36**

* Gut luminal metabolites with significant changes (fold change >2, $p < 0.05$) after 7 days of
treatment with tacrolimus (TAC), prednisone (PRED), mycophenolate mofetil (MMF), or FTY720
compared to untreated controls. Numbers in bold indicate total metabolites significantly
increased (↑) or decreased (↓) for each drug. Key metabolic pathways affected are listed for
each drug. SCFAs: short-chain fatty acids; TMA: trimethylamine. Data derived from untargeted
metabolomics analysis using capillary electrophoresis-mass spectrometry.

Table 2. Immunosuppression drugs and dosage.

Treatment Groups	Dose*(mg/kg/day)	Source (catalog #)	Route
Tacrolimus	3 mg/kg/day	Sigma Aldrich (Y0001926)	Subcutaneous [92]
Prednisolone	5 mg/kg/day	Sigma Aldrich (P6004)	Intraperitoneal [59]
Mycophenolate mofetil	30 mg/kg/day	Sigma Aldrich (SML0284)	Intraperitoneal [93]
FTY720	3 mg/kg/day	Sigma Aldrich (SML0700)	Oral [94]

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 Figure legends

 Figure 1: Immunosuppressants dynamically influence gut microbiome. Mean

 pairwise distance comparison between conditions using gut microbiome **A)** taxonomic

 profiles and **B)** functional pathway profiles. Both taxonomic and functional pathways

 profiles are relative abundance of taxonomic groups. Distance calculated based on

 average Bray-Curtis distance per group. **C**) Principal Components Analysis of gut

 microbiome. Robust centered log ratio transformation applied on count compositional

 data, PCA calculated based on Bray-Curtis distance. **D)** Differentially abundant

 taxonomic groups. No treatment control group for comparison. Pairwise comparisons

 between the relative abundance of taxonomic groups and no treatment control. Fold

 changes in taxonomic group abundance expressed using base 2 logarithmic ratio.

 Figure 2: Immunosuppressants distinctively influence gut metabolism. Volcano

 plots combine results from fold change (FC) analysis to show significantly increased

 metabolites after 7 days for **A**) TAC, **B**) PRED, **C**) MMF, or **D**) FTY720 treatment

 compared to no treatment control. A metabolite is shown if FC is >2 and p value is

 <0.05 based on 2-sample t-tests. Original metabolite measurement without

 normalization in FC analysis.

 Figure 3: Immunosuppressants temporally shift intestinal immune responses. A)

 Bar plots showing the number of differentially expressed genes (DEGs) after 7 and 30

 days of treatment compared to the control group. Yellow and blue bars indicate the

 number of upregulated genes, while red and green bars represent the number of

1046 downregulated genes. **B**) Dot plot illustrating over-represented immune-related

1047 functional pathways of DEGs in small intestinal tissues across different treatments

1048 compared to the control group, analyzed at 7 and 30 days of treatment. Gene Ontology

1049 pathways used. Dot size represents the -log10(p-value), indicating the statistical

1050 significance of pathway enrichment.

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Figure 4: Effects of TAC, PRED, MMF, and FTY720 on mLN cell content, cell

distribution, and structure. Flow cytometry analysis of **A)** B220+ B cells and CD4+ T

cells on day 30 after PRED treatment and **B)** Foxp3+ Tregs on day 30 after FTY

treatment in mLNs. **C)** Heatmaps depict changes in expression with TAC, PRED, MMF,

and FTY relative to control in mLN after 3, 7, and 30 days of treatment. IHC of **D)** CR

and **E)** HEV Foxp3+ Tregs on days 3, 7, and 30 of mLN. IHC of **F)** CR and **G)** HEV la4:

la5 on days 3, 7, and 30 of mLN. Barplot showing detailed plots at each timepoint are in

Supplemental Figure 6. Heatmaps depict changes in expression with tacrolimus,

prednisolone, MMF, and FTY720 relative to control in **H)** CR and **I)** around HEV in mLN.

3 mice/group. 2-3 LNs/mouse, 2-3 sections/LN, 7-30 fields/tissue. One-way ANOVA. * p

< 0.05; ** p < 0.01, *** p < 0.001, **** p < 0.0001.

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Figure 5: Effects of TAC, PRED, MMF, and FTY720 on pLN cell content, cell

distribution, and structure. Flow cytometry analysis of **A)** CD4+ T cells on day 30

after MMF treatment, **B)** B220+ B cells and CD4+ T cells on day 7 and **C)** CD4+ T cells

and Foxp3+ Tregs after FTY treatment in pLNs. **D)** Heatmaps depict changes in

expression with TAC, PRED, MMF, and FTY relative to control in pLN after 3, 7, and 30

days of treatment. IHC of **E)** CR and **F)** HEV Foxp3+ Tregs on days 3, 7, and 30 of pLN.

IHC of **G)** CR and **H)** la4: la5 on days 3, 7, and 30 of pLN. Barplot showing detailed

plots at each timepoint are in **Supplemental Figure 7.** Heatmaps depict changes in

expression with tacrolimus, prednisolone, MMF, and FTY720 relative to control in **I)** CR

and **J)** around HEV in pLN. 3 mice/group. 2-3 LNs/mouse, 2-3 sections/LN, 7-30

fields/tissue. One-way ANOVA. * $p < 0.05$; ** $p < 0.01$, *** $p < 0.001$, **** $p < 0.0001$.

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Figure 6. MMF and FTY720 induce a rapid anti-inflammatory response during

allogeneic stimulation. IHC for Foxp3+ Tregs in the **A**) pLN and **C**) mLN. La4:La5 in **B**)

pLN and **D**) mLN. Heatmap summarizes significant changes in expression comparing

allogeneic stimulation plus MMF or FTY720 to allogeneic stimulation alone in **E**) pLN

and **F**) mLN. 3 mice/group. 2-3 LNs/mouse, 2-3 sections/LN, 7-30 fields/tissue. One-

way ANOVA: * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$, **** $p < 0.0001$.

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Figure 7. MMF and FTY decrease the La4:La5 in FRC-Lama5-KO mLN. IHC showing

A, D) laminin α4; **B, E**) laminin α5; and **C, F**) La4:La5 in mLN of WT, FRC-Lama4-KO

mice (**A-C**) and FRC-Lama5-KO mice (**D-F**) on day 3. **G**) Heatmap summarizes

significant changes relative to control. 3-5 mice/group. 2-3 LNs/mouse, 2-3 sections/LN,

7-30 fields/tissue. One-way ANOVA: * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$, **** $p <$

0.0001.

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Figure 8. MMF increases the La4:La5 in FRC-Lama5-KO, and FTY decreases the

La4:La5 in FRC-Lama4-KO pLN. IHC for **A, D**) laminin α4; **B, E**) laminin α5; and **C, F**)

La4:La5 in pLN WT, FRC-Lama4-KO mice (**A-C**) and FRC-Lama5-KO mice (**D-F**) on

day 3. **G**) Heatmap summarizes significant changes relative to control. 3-5 mice/group.

2-3 LNs/mouse, 2-3 sections/LN, 7-30 fields/tissue. One-way ANOVA* p < 0.05, ** p <

0.01, *** p < 0.001, **** p < 0.0001.

Figure 1.

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1099   **Figure 2.**

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1102 **Figure 3.**
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