**Epithelial Hexokinase 2 expression is a marker for intestinal inflammation**

Saskia Weber-Stiehl[1*], Jan Taubenheim[2*], Lea Järke[1*], Christoph Röcken[3], Stefan Schreiber[4],

Konrad Aden[1,4], Christoph Kaleta[2], Philip Rosenstiel[1], Felix Sommer[1]

[1] Institute of Clinical Molecular Biology, University of Kiel, Rosalind-Franklin-Straße 12,

24105 Kiel, Germany

[2] Institute of Experimental Medicine, University of Kiel, Michaelisstr. 5, 24105 Kiel, Germany

[3] Department of Pathology, University Hospital Schleswig-Holstein, Campus Kiel, Arnold-

Heller-Straße 3/House U33, 24105 Kiel, Germany

[4] Department of Internal Medicine I, University Hospital Schleswig-Holstein, Campus Kiel,

24105 Kiel, Germany

[*] These authors contributed equally with shared first authorship.

**Correspondence:** Felix Sommer, Institute of Clinical Molecular Biology (IKMB), Christian-

Albrechts-University (CAU) Kiel, Rosalind-Franklin-Str. 12, Kiel D-24105, Germany; tel: +49(0)

431/500-15146; email: f.sommer@ikmb.uni-kiel.de

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**Abstract**

**Background:** Inflammation is characterized by a metabolic switch promoting glycolysis and lactate production. Hexokinases (HK) catalyze the first reaction of glycolysis and inhibition of epithelial HK2 protected from colitis in mice. HK2 expression has been described as elevated in patients with intestinal inflammation, however there is conflicting data from few cohorts especially with severely inflamed individuals, thus systematic studies linking disease activity with HK2 levels are needed.

**Methods:** We therefore examined the relationship between HK2 expression and inflammation severity using transcriptome data derived from the mucosa of thoroughly phenotyped patients suffering from intestinal inflammation. The mucosa undergoes a cellular remodeling during inflammation due to an infiltration of HK2-negative immune cells and a loss of the apical epithelium – the main site of HK2 expression.

**Results:** Normalizing for the epithelial loss clearly identified epithelial HK2 expression as gradually increasing with disease activity and remaining elevated at high inflammation scores. We then performed immunofluorescence analyses on colonic biopsies of unrelated patients with intestinal inflammation and confirmed the RNA-based findings on protein level. HK2 protein expression was mostly restricted to brush border enterocytes and these cells along with HK2 levels vanished with increasing disease severity.

**Conclusions:** Our findings clearly define dysregulated epithelial HK2 expression as an indicator of disease activity in intestinal inflammation and suggest targeted HK2-inhibition as a potential therapeutic avenue.

## Introduction

Prevalence of inflammatory bowel disease (IBD) is rising globally causing severe health issues and drastically reducing quality of life. These diseases are multifactorial with complex interactions of multiple genetic and environmental factors[1]. Despite decades of intense research, the exact etiology of IBD remains mostly unknown limiting treatment options[2]. In recent years it became evident that ongoing inflammation is characterized by changes in metabolic activity, in particular glucose metabolism[3–5]. Hexokinases (HK) catalyze the first and irreversible step of glycolysis – the phosphorylation of glucose to glucose-6-phosphate – and are therefore crucial for glucose metabolism and maintenance of homeostatic body functions. In mammals, five HK isoenzymes have been identified: HK1, HK2, HK3, GCK (Glucokinase) and HKDC1 (HK domain containing 1). Each of these HK isoenzymes display specific tissue expression patterns and glucose affinities. HK2 is of particular interest as it is the most abundant HK member within the intestine, exerts high HK activity, responds to various internal as well as external stimuli and shows elevated levels during inflammation[6]. Recently, we further revealed an upregulated expression of HK2 specifically in inflamed compared to non-inflamed tissue of the same patient, irrespective of the type of intestinal inflammation and demonstrated that ablation of HK2 in intestinal epithelial cells (IEC) protects from acute intestinal inflammation, suppresses cell death and alters mitochondrial function[6]. These findings placed HK2 as a molecular target to treat intestinal inflammation. However, when expanding our research about the role of HK2 in intestinal inflammation, we encountered transcriptome datasets with contradictory findings of either an unaltered HK2 expression or even downregulation, especially in severely inflamed patients undergoing bowel resection[7]. We therefore aimed to investigate the cause of this apparent contradiction with the goal to clarify the relation between HK2 expression and intestinal inflammation. Here, we show that with increasing inflammation severity the intestinal mucosa is gradually remodeled, which comprises a partial loss of the apical epithelium – the primary source of HK2 expression – and a simultaneous infiltration of immune cells. Normalizing for this cellular remodeling clearly demonstrates a gradual upregulation of HK2 expression with severity of intestinal inflammation.

**Results**

**Epithelial HK2 expression increases with high inflammation scores**

To determine a possible correlation of HK2 expression and disease activity, we analyzed RNA

sequencing (RNA-seq) data derived from intestinal biopsies of patients suffering from IBD of

two independent clinical studies[8,9], for which accompanying disease activity scores for each

sample were available. This analysis revealed that HK2 expression initially increased gradually

reaching peak expression at mid disease activity (inflammation score ~ 0.4) and then declined

with high inflammation scores (Figure 1A). As within the intestine HK2 is mainly expressed by

epithelial cells of the apical mucosa (Figure S1), we hypothesized that with very high levels of

inflammation these apical epithelial cells and therefore also the main site of HK2 expression

might be lost due to shedding and apoptosis or transdifferentiation resulting in decreased

HK2 expression levels in bulk RNA-seq data. To investigate this hypothesis, we assessed the

expression levels of three different epithelial marker genes, namely E-cadherin (*ECAD*),

epithelial cell adhesion molecule (*EPCAM*) and Villin1 (*VIL1*). Supporting our hypothesis, the

expression levels of these epithelial markers were all declining with increasing inflammation

severity (Figure 1B-D), which indicates progressive epithelial damage caused by the

inflammation. To account for this epithelial loss, we then normalized the expression of HK2

to that of these epithelial markers, thereby yielding an epithelial HK2 expression. Importantly,

epithelial HK2 expression gradually increased in correlation to disease activity and then

remained elevated at high inflammation scores (Figure 1E). To test whether the observed

expression changes were unique to the combined clinical cohorts, we also performed these

analyses separately for each clinical cohort (Figure S2). Importantly, we observed for both

clinical cohorts the same changes in *HK2*, *ECAD*, *EPCAM*, *ECAD* and epithelial HK2 expression.

We also tested for potential differences between the two main types of IBD, Crohn´s disease

(CD) and ulcerative colitis (UC) and found similar responses both for CD and UC (Figure S3).

Finally, correlation of HK2 expression with disease severity was confirmed using linear

regression models (Figure S4 and Table S1).

**Immune cell infiltration during intestinal inflammation**

Another source for cellular remodeling during inflammation is tissue infiltration by immune

cells. Using the bulk RNA-seq data we also investigated the proportions of various immune

cell types in relation to disease severity. First, we performed a candidate gene-based analysis

and chose several immune marker genes to assess infiltration of T cells, T helper 1 (Th1) cells,

T helper 2 (Th2) cells, B cells, basophils, neutrophils, eosinophils, and macrophages (Figure

2A). In contrast to *HK2* and the epithelial marker genes, the expression of these individual

immune cell genes increased with the inflammation score. However, slight alterations in the

expression patterns could be observed, mostly depending on the immune cell type.

Expression of the T cell marker genes *TRAC*, *CD3D* and *CD3E,* the B cell marker gene *CD19,* the

eosinophil marker gene *CCR2*, the neutrophil marker genes *CD14* and *CXCR4* as well as the

basophil marker genes *ITGA2B* and *ITGA4* all gradually increased in a linear fashion with the

inflammation score (Figure 2). In contrast, the marker genes for Th2 (*IL4*, *IL5*, *IL13*), Th1 (*IFNG*,

*IL12*) cells and macrophages (*ITGAM*, *IL1B*) all only displayed biphasic expression patterns

with a first phase characterized by slight increases in expression until an inflammation score

of approximately 0.5 and a second phase at higher disease severity characterized by larger

changes in their expression. The greatest expression changes were detected for the

macrophage marker genes *ITGAM* and *IL1B* suggesting the greatest relative increase of these

cells with increasing disease activity. Normalizing *HK2* expression to the candidate immune

cell genes revealed gradual declines with increasing inflammation scores for all tested

immune cell types indicating that the immune cell infiltration leads to more cells present in

the mucosa, which do not or only express very little *HK2* (Figure 2B).

Next, we moved from the candidate gene-based to a systematic cellular deconvolution of the

bulk RNA-seq data using MuSiC[10]. This program uses a reference single cell data set and cell

type-specific expression profiles to derive the abundance of the individual cell types from bulk

RNA-seq data. We restricted this analysis to the eight most abundant cell types. The

proportion of epithelial cells gradually declined with increasing inflammation scores whereas

the proportion of macrophages, B cells and regulatory T cells increased (Figure 3A). The

proportion of mast, cytotoxic T cells and dendritic cells remained mostly unchanged.

Therefore, this data indicates a massive gradual cellular remodeling of the intestinal mucosa

during inflammation and the findings of the systematic cellular deconvolution supported

those of the candidate gene-based analyses. Furthermore, the drastic decrease in the

proportion of epithelial cells with increasing inflammation scores demonstrates again that the
main source of *HK2* expression is lost during the disease course. This is therefore also
reflected by the outcome that after accounting for the abundance of epithelial cells the
normalized epithelial *HK2* expression gradually increased with disease severity (Figure 3B).

**Apical HK2 expression in mature enterocytes correlates with inflammation severity**

To validate the transcriptome data and to elucidate the biogeography of the changes in tissue
architecture in relation to HK2 expression, we performed immunostainings on colonic
biopsies of an independent set of UC patients that had been thoroughly scored for histological
disease severity (Nancy score). HK2 protein levels increased non-significantly in mildly
inflamed (score 1) compared to non-inflamed specimen (score 0) and remained unaltered at
medium or strong inflammation (scores 2 and 4) (Figure 4A,C). In contrast, protein levels of
the epithelial marker E-cadherin (ECAD) were unchanged in mild (score 1) but non-
significantly reduced at medium or strong inflammation (scores 2 and 4) compared to non-
inflamed control samples (Figure 4A,C). Notably, structural epithelial damage of inflamed
biopsies was detectable with increasing inflammatory scores (Figure 4C) with loss of epithelial
cells and an infiltration of other cells, probably immune cells, into the inflamed mucosa. In
addition, many epithelial cells that were not shed but still present at the inflamed site had
reduced ECAD expression, which could indicate transdifferention of these inflamed apical
enterocytes. Normalizing the HK2 protein levels to those of ECAD revealed that epithelial HK2
protein levels (Figure 4B) significantly increased with inflammation regardless of disease
severity (scores 1-4). This epithelial HK2 protein expression in inflamed intestinal mucosal
biopsies mirrored and confirmed the transcript patterns of the RNA sequencing analysis.

Furthermore, we used recently-published data from single cell RNA sequencing of mucosal
biopsies from IBD patients[11] to investigate the *HK2* expression on a cellular level in relation to
intestinal inflammation. This analysis of single cell RNA sequencing data confirmed our
findings from the bulk RNA sequencing and immunofluorescence, as *HK2* was mainly
expressed by mature enterocytes and then followed by goblet cells and immature
enterocytes (Figure 5A,B). Next, we checked for *HK2* expression pattern in enterocytes at
different inflammatory stages (healthy vs. inflamed vs. non-inflamed) and found that the
lowest *HK2* expression can be observed in healthy controls, while cells from inflamed samples

showed an increased HK2 expression (Figure 5C). Non-inflamed samples had intermediate

*HK2* expression levels (Figure 5C), which indicates that HK2 expression is dysregulated in

enterocytes of IBD patients also in the absence of an overt inflammation. This pattern of *HK2*

expression was also present although less predominant in immature enterocytes and goblet

cells. Finally, we looked into the number of these three cell types that are detected in the

three health states and found that mature and immature enterocytes are lost during

inflammation, while incomplete recovery can be observed in non-inflamed tissue of IBD

patients (Figure 5D). Altogether, the single cell data therefore supports a cellular

reprogramming during inflammation with loss of mature enterocytes (the main cell type of

the apical epithelium) and a dysregulated *HK2* expression in inflamed apical epithelial cells.

In summary, both the RNA and protein data clearly demonstrate that epithelial HK2 was

overexpressed during intestinal inflammation, which highlights HK2 as an indicator of active

disease. Building upon previous findings from murine models that inhibition of HK2

expression either genetically or via the microbial metabolite butyrate protect from

experimental colitis[6], these new data now show that indeed HK2 expression is dysregulated

in the mucosa of patients with active inflammation and therefore suggest that targeting HK2

may represent a promising approach to suppress intestinal inflammation in humans.

**Discussion**

There is a growing need for valid disease biomarkers and molecular targets for intestinal inflammation as prevalence is increasing and despite significant efforts the cause of these diseases remains unknown. In a recent study, HK2 was identified as upregulated under inflammatory conditions in mice and humans irrespective of disease subtype (CD, UC or non-IBD colitis) and suppressing HK2 expression even protected from experimental colitis[6] placing HK2 as a potential disease marker. However, we found other transcriptome datasets generated from severely inflamed patients in which HK2 expression was unaltered or even reduced[7]. Here, we therefore thoroughly investigated the relationship between HK2 expression and disease severity by analyzing transcriptome data derived from intestinal biopsies of patients with intestinal inflammation for which corresponding data on disease severity (HBI, Mayo score, Nancy score) were available. In addition, we performed cellular deconvolution analyses of the RNA sequencing data to infer the changes in cell proportions during inflammation. Finally, we used immunofluorescence staining to clarify the changes in HK2 protein biogeography in the mucosa during inflammation. Using these approaches, we demonstrated that raw HK2 RNA and protein expression at first gradually increased with the inflammation scores, yet after reaching a critical inflammation score HK2 expression declined again at very severe inflammation. These findings integrate the seemingly conflicting data from previous studies[7,12–14] by linking HK2 expression to disease severity. We further were able to demonstrate that loss of HK2 expression at high inflammation levels was dependent on the disruption of brush border enterocytes as both the expression of epithelial marker genes were decreasing with the inflammation score and immunofluorescence analyses clearly pointed to a destruction of the apical epithelium – the main site of HK2 expression[15,16]. These findings are in line with clinical practice, in which epithelial damage is a key criterion to classify an increasing inflammatory state during IBD[17], for example with epithelial erosion being a feature of histological inflammation in the Nancy score. By analyzing single cell RNA sequencing data derived from an independent set of UC patients, we confirmed that in the gut epithelial cells are the main source of *HK2* expression, both in terms of per cell expression as well as the number of cells contributing to overall expression (Figure 5). These cells are consequently lost due to increasing inflammation in the single cell data set, which would explain the decrease of *HK2* expression at high levels of inflammation[11]. Finally, we were able

to deconstruct the simultaneous epithelial erosion and immune cell infiltration into the submucosa during intestinal inflammation from the RNA sequencing data. Based on the overall abundance and their induction (fold change) during inflammation, infiltrating macrophages seemed most relevant, but also other cell types such as B cells increased in proportion. Especially the expansion of macrophages could be important and contributing to the loss of epithelial cells and therefore HK2 expression during inflammation as IL-1β interferes with the tight junction complexes between IECs[11–13] and thereby increases intestinal permeability[21]. Similar to epithelial erosion, immune cell infiltration is a classical feature of inflammation and used in clinical practice to histologically evaluate disease severity, in particular as a feature of the Nancy score[22]. In addition, transdifferentiation of HK2-positive mature enterocytes into other cell types such as HK2-negative immature enterocytes and stem cells[23,24] or maybe even into mesenchymal cells as during epithelial–mesenchymal transition [25] could also contribute to a reduction of HK2 levels in the inflamed mucosa. In summary, our findings imply that during the course of inflammation, changes in the cellular composition of the mucosa, i.e. loss of brush border epithelial cells expressing high levels of HK2 and a corresponding massive recruitment of immune cells expressing little to no HK2 to the site of inflammation, affect the overall bulk HK2 expression leading to an overall reduction despite higher expression levels in fewer and fewer HK2-positive cells (Figure 6). Overall, we want to highlight the importance of taking into consideration the biogeography and changes in cell type expression ratios, when trying to identify disease biomarkers, especially regarding complex diseases such as intestinal inflammation involving various internal and external factors. Most previous studies used bulk expression data in the search for disease biomarkers or therapeutic targets and therefore may have missed other locally restricted but disease relevant genes like *HK2*. In our analyses HK2 expression was only significantly associated with disease severity after considering epithelial cell abundance (Figure 1E and Figure 3B) and we speculate that this pattern will also hold true for other genes with an apical or otherwise restricted expression. For example, other hexokinases (HK1, HKDC1), proteins responsible for the uptake of dietary nutrients (e.g. SGLT1, SLC15A1) or those contributing to the intestinal epithelial barrier (e.g. CLDN15, ZO-1) also show a predominant or even restricted apical expression and therefore possibly have not yet been identified as disease biomarker during severe intestinal inflammation due to the loss of their primary expression site. Notably, epithelial remodeling not only includes the partial loss of

apical enterocytes and infiltration of lymphocytes, but also includes hyperregeneration, metaplasia and a loss of goblet cells. The advent of single cell technologies such as single cell RNA sequencing promise to pave the way for more refined analyses that will enable the discovery of more suitable disease markers and to enhance our understanding of the molecular and cellular mechanisms during disease progression.

**Conclusions**

HK2 has not yet been described as IBD risk gene, most probably due to its site-restricted expression, but here we were able to show an association of HK2 and epithelial status during intestinal inflammation. Therefore, epithelial HK2 expression correlates with disease severity making it a useful indicator of intestinal inflammation and highlighting the therapeutic potential of targeting HK2. However ultimately, clinical studies are required to indeed demonstrate the feasibility and efficacy of a HK2-targeted intervention in IBD patients.

**Materials and methods**

**Human studies**

We used mucosal transcriptomic data and fixed intestinal specimen of two independent large longitudinal clinical studies, namely the EMED[9] (191 samples) and FUTURE[8] (86 samples) cohorts (trial IDs: EudraCT number 2016-000205-36 and ClinicalTrials.gov NCT02694588), for which accompanying disease activity scores (Harvey-Bradshaw Index (HBI)/Mayo Score) were available. The HBI[26] is used to quantify disease activity in patients suffering from Crohn's disease (CD), one of the two IBD subtypes. Here, variables such as general well-being, abdominal pain and abdominal mass with each having scores of 0-3, as well as the number of liquid stools per day (1 point per stool) and other complications (1 point per complication) yield the "open-ended" total score (Table S2). The Mayo score is used for ulcerative colitis (UC), the other clinical IBD subtype. This scoring system accounts for general well-being, rectal bleeding, endoscopic results and stool frequency with scores of 0-3 per category for an overall score ranging from 0 to 12 (Table S2). To facilitate a direct comparison of both scores despite them being "open-ended" and "discrete", we calculated a general inflammation score by setting the highest score in the dataset for each HBI and Mayo score to 1 and scaled the score of each patient accordingly (see Table S2).

**Transcriptome data analysis**

RNA sequencing data derived from mucosal biopsies of patients with various grades of intestinal inflammation were retrieved from two previously published IBD cohorts[8,9]. Read counts were transformed to Transcripts Per Million (TPM) values to normalize for differential sequencing depths among samples. TPM data were then plotted per sample against the inflammation score using R (version 4.2.2) and ggplot2 (3.4.3). Trendlines were calculated including all data points and using the distance to the actual location as a weight to enable a robust calculation and avoiding overfitting especially in areas of scarce sampling. To test for an association of gene expression to changes in the inflammation score we used variance-stabilization transformation of raw read counts and fitted linear mixed models with following form: vst(geneExpression) ~ Age + Gender + Diagnosis + InflammationScore. Statistical analyses were performed using R (version 4.2.2[27] with the following packages: lm4 (version

284   1.1-29)[28], lmerTest (version 3.1-3)[29], car (version 3.0-13)[30], ggplot2 (version 3.3.6)[31] and

285   DESeq2 (version 1.38.3)[32].

**Cellular deconvolution**

We deconvoluted the bulk RNA data using the MuSiC package (version 1.0.0)[10] with default

values. As reference, we used a single cell data set from UC patients described in a study by

Smillie et. al.[11]. The original single cell data set was split in "epithelial", "immunogenic" and

"stromal" subsets, which we rejoined in our analysis. Due to limited resolution in

deconvolution approaches and to reduce cell diversity, we pooled ontogenetically closely

related cell types (Table S3). We restricted the analysis to the eight most abundant cell types.

**Single cell RNA sequencing**

Single cell data was obtained from the Single Cell Portal (accession SCP259) initially described

by Smillie and colleagues[11]. The different cell populations were rejoined and general quality

controls were performed. In short, we filtered cells with read counts below 1000 (low quality).

Further, we removed cells with less than 200 or with more than 2720 (two-fold of the

standard deviation of expressed genes across all cells) differently expressed genes to account

for spurious sequencing depth and removal of duplicates. All cells with a mitochondrial RNA

content above 5% were also removed. For plotting, gene expression values where log-

transformed. Plotting and quality filtering was performed in R (version 4.2.2) using ggplot2

(version 3.4.3) and Seurat (version 4.4.0)[33].

**Immunofluorescence**

5 µm sections of paraffin-embedded colonic biopsies were deparaffinized with Xylol

substitute (Roth), incubated in citrate buffer for 3 minutes and subsequently blocked in 5 %

BSA-PBS and 0,2 % TritonX for 30 minutes. Primary anti-E-cadherin (1:400 in 1 % BSA, #3195,

Cell Signaling Technology) and anti-Hexokinase 2 (1:500 in 1% BSA, Cat# NBP2-02272, Novus

Biologicals, Colorado, US) antibodies were incubated over-night. Sections were washed,

incubated with secondary antibodies (Alexa Fluor 488 goat anti mouse, Invitrogen, A32731

and Alexa Fluor 555 goat anti rabbit, Invitrogen, A21430) for 45 minutes at room

temperature. DAPI (1:40000 in PBS, D9542, Sigma Aldrich, St. Louis, US) was used for DNA

counterstaining. Slides were mounted using antifade mounting media (DAKO, Hovedstaden,

Denmark). The quantitative analysis was performed using fluorescence microscopy for

stained samples using the imager Z1 microscope (ZEISS, Jena, Germany) and ZEN software

(version 3.0). Images were taken by a digital camera system (AxioCam HrC/HrM, Zeiss, Jena,

Germany) and ApoTome (ZEISS, Jena, Germany). Fluorescence signal intensity was measured

using Fiji/ImageJ software.

**Statistics**

General statistical analyses were performed using the GraphPad Prism 9 (GraphPad Software

Inc., La Jolla, USA). For pairwise comparisons the Mann-Whitney-U test was used whereas for

multiple comparisons one-way ANOVA with false discovery rate (FDR) correction were

performed. Data are shown as mean ± standard error of the mean (SEM). A p-value of ≤ 0.05

was considered as significant (*). A p-value of ≤ 0.01 was considered as strongly significant

(**) and p-value of ≤ 0.001 as highly significant (***). For longitudinal data, linear mixed effect

models were fitted with where gene expression was used as dependent variable, while

inflammation score and patient ID were fixed and random factors. The models were fitted

using lme4 (version 1.1-31) and statistical testing for coefficients was performed using

lmerTest (version 3.1-3). Model validity was tested against a null-model using log-likelihood

ratio tests and model assumption were evaluated by diagnostic plotting of model residuals,

data point influence strength and random factor fitting.

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**Declarations**

*Ethics approval and consent to participate*

All research complied with relevant ethical regulations, and usage of human tissue material and transcriptome data was approved by the ethics committee of the Medical Faculty at Kiel University (A156/03, A124/14, TJ301).

*Consent for publication*

Not applicable

*Availability of data and material*

Not applicable. All data was retrieved (see method section).

*Competing interests*

SS reports indirect stock ownership in Gerion Biotech GmbH as well as consulting and personal fees from AbbVie, Allergosan, Amgen, Arena, BMS, Biogen, Celltrion, Celgene, Falk, Ferring, Fresenius, Galapagos/Gilead, HIKMA, I-Mab, Janssen, Lilly, Morphic, MSD, Mylan, Pfizer, Prometheus, Protagonist, Provention Bio, Sandoz/Hexal, Takeda, and Theravance. PR reports stock ownership in Gerion Biotech GmbH and consulting fees from Takeda.

All other authors declare no competing interests.

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*Authors' contributions*

SW, LJ and FS designed research. SW, JT, LJ and FS performed experiments and analyzed the data. SS, CR, KA and PR provided samples and RNAseq data. SW, JT, LJ and FS prepared the figures. KA, PR, CK and FS obtained funding. SW, JT, LJ and FS co-wrote the manuscript with critical input from all authors. All authors read and approved the final manuscript.

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**Figures and tables**

**Figure 1: Epithelial HK2 expression correlates with inflammation severity. A-D)** Expression

(TPM, transcript per million) of *HK2* **(A)** and the epithelial marker genes *ECAD* **(B)**, *VIL1* **(C)** and

*EPCAM* **(D)** in the intestinal mucosa of patients with various degrees of gut inflammation.

Note that at high inflammation scores (>0.4) expression of *HK2* and the epithelial marker

genes all decrease indicating epithelial erosion. The inflammation score was calculated as a

scaled Harvey-Bradshaw index (Crohn's disease) or Mayo score (ulcerative colitis) to

accommodate both disease types. **E)** *HK2* expression increases with inflammation scores after

normalization to epithelial marker gene expression. The red lines represent the mean

expression trendline. The number in the upper right corner represents the p value for the

correlation between gene expression and inflammation score as determined by linear mixed

model.

**Figure 2: Expression of candidate immune marker genes increases with inflammation severity.** Expression of selected marker genes, which are characteristic for individual immune cell types, were evaluated in patient biopsies. Expression of virtually all candidate immune marker genes increase with inflammation severity indicating an expansion of immune cells – a known feature of intestinal inflammation.

**Figure 3: Deconvolution of mucosal transcriptomes demonstrates cellular remodeling during inflammation. A)** MuSiC cell deconvolution package was used to estimate ratios of the most abundant cell types based on the mucosal patient transcriptomes. Notably, epithelial cell abundance is decreasing with inflammation, whereas abundances of macrophages and B cells, the two most frequent immune cell types that combine for approximately 35 % proportion, are increasing with inflammation severity. Therefore, cell deconvolution supports the candidate gene analyses indicating mucosal cellular remodeling characterized by a reduction of epithelial cells and an expansion of immune cells with increasing inflammation severity. **B)** *HK2* expression increases with inflammation scores after normalization to epithelial cell abundance.

**Figure 4: Epithelial HK2 protein expression increases with inflammation severity.** Colonic mucosa biopsies of unrelated UC patients were stained for HK2, epithelial cells (ECAD) and nuclei (DAPI). All biopsies were analyzed for classical histological signs of inflammation using the Nancy Score (see method section). **A)** Quantification of HK2 and ECAD signal intensity in apical intestinal epithelium. Note that HK2 and ECAD level decrease with higher Nancy Scores. **B)** HK2 protein expression normalized to ECAD levels. Note that after epithelial normalization HK2 expression increases with inflammation severity, i.e. Nancy Score. Significance testing was performed using one-way ANOVA compared to Score = 0. **C)** Representative images of the multiplex immunofluorescence stainings (HK2, ECAD, DAPI). Scale bar indicates 50 μm. n = 7 – 10 per group. Note that microscopic images were taken from areas with rather intact epithelium for a better visualization of the mucosal structure.

**Figure 5: *HK2* is mainly expressed by mature enterocytes and increased during inflammation.** *HK2* expression was analyzed in published single cell RNA sequencing data derived from mucosal biopsies of UC patients[11]. **A)** *HK2* expression per cell type. **B)** Abundances of cell types expressing *HK2*. **C)** *HK2* expression in mature & immature enterocytes and goblet cells divided into samples from healthy controls, inflamed and non-inflamed mucosa. **D)** Abundances of cell types expressing *HK2* after stratification into disease group.

**Figure 6: Model of HK2 expression changes in relation to intestinal inflammation severity.** HK2 is predominantly expressed by apical epithelial cells and HK2 expression in these cells increases with disease severity. However, during the course of inflammation the cellular composition of the mucosa changes. In severely inflamed tissue, the epithelium is disrupted, brush border epithelial cells with high levels of HK2 are shed and lost, whereas immune cells with little to no HK2 expression are recruited. Therefore, overall HK2 expression is reduced in bulk tissue samples under severe inflammation despite local high expression in remaining HK2-positive epithelial cells. Created with BioRender.com

**Figure S1: HK2 expression in the intestinal mucosa.** Biopsies from small and large intestine were immuno-stained to localize the HK2 protein in the mucosa. Note that HK2 expression is mainly confined to epithelial cells of the apical mucosa both in the small and large intestine.

**Figure S2: Epithelial *HK2* expression correlates with disease severity in both clinical cohorts.**

Expression data (TPM, transcript per million) of *HK2* and the epithelial marker genes *ECAD,*

*VIL1* and *EPCAM* in the intestinal mucosa of patients with various degrees of gut inflammation

split per clinical cohort FUTURE and EMED.

**Figure S3: Epithelial *HK2* expression correlates with inflammation severity regardless of disease subtype.** Expression data (TPM, transcript per million) of *HK2* and the epithelial marker genes *ECAD, VIL1* and *EPCAM* in the intestinal mucosa of patients with various degrees of gut inflammation split per disease subtype CD and UC.

**Figure S4: Correlation analysis for HK2 expression and inflammation score.** Three different linear mixed effect models (PatientID as random factor) were fitted to the data. First, all data points were used to estimate the overall effect of *HK2* expression and inflammation score. Afterwards the data set was split into "low scores" (inflammation score < 0.5), "high scores" (inflammation score >0.5) and fitted models for these data subsets. While the first model showed no significant association between *HK2* expression and inflammation, there is a positive association at "low scores" and a negative association at "high scores" (see also table S1) indicating that *HK2* expression increases with inflammation during lower disease scores and then decreases at higher disease scores.

 **Table S1: Comparison of linear mixed model statistics for the coefficients explaining HK2**

540   **expression with inflammation scores for models created with all data, low (score <0.5) or**

541   **high (score >0.05) inflammation scores only.**

| | Estimate | Std. Error | DF | t-value | p-value | |
|---|---|---|---|---|---|---|
| **All data** | 44.73 | 32.07 | 179.94 | 1.395 | 0.165 | |
| **Low score** | 195.86 | 59.48 | 154.44 | 3.293 | 0.0123 | ** |
| **High score** | -178.6 | 78 | 114.87 | -2.29 | 0.0239 | * |

543 **Table S2: Translation of the HBI and Mayo Score into a general Inflammation Score.** The

544 definitions of the HBI and Mayo scores are listed. The Inflammation Score was calculated by

545 scaling each the HBI and Mayo Score from 0-1 and then merging the scores.

| HBI | | |
|---|---|---|
| **Criterium** | **Score** | |
| General well-being | Very well | 0 |
| | Slightly poor | 1 |
| | Poor | 2 |
| | Very poor | 3 |
| | Terrible | 4 |
| Abdominal pain | None | 0 |
| | Mild | 1 |
| | Moderate | 2 |
| | Severe | 3 |
| Abdominal mass | None | 0 |
| | Dubious | 1 |
| | Definite | 2 |
| | Definite with tenderness | 3 |
| Stool frequency | Number of liquid stools | 1 for each lquid stool per day |
| Complications | Arthralgia, Uveitis, Erythema nodosum, Apthous ulcers, Pyoderma gangrenosum, Anal fissure, New fistula, Abscess | 1 point each |

| Mayo Score | | |
|---|---|---|
| **Criterium** | **Score** | |
| Number of stools per day | Normal for the patient | 0 |
| | 1-2 stools more than normal | 1 |
| | 3-4 stools more than normal | 2 |
| | 5+ stools more than normal | 3 |
| Rectal blood | No blood | 0 |
| | Blood streaks in less the half of evacuations | 1 |
| | Evidence of fresh blood in most of the evacuations | 2 |
| | Bowel movements with fresh blood | 3 |
| Endoscopy | Normal or inactive disease | 0 |
| | Mild disease (erythema, decreased vascular pattern, mild friability) | 1 |
| | Moderate disease (marked erythema, lack of vascular pattern, friability, erosions) | 2 |
| | Severe disease (spontaneous bleeding, ulceration) | 3 |
| Global medical asessment | Normal | 0 |
| | Mild disease | 1 |
| | Moderate disease | 2 |
| | Severe disease | 3 |

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| Calculation of inflammation score | | | | | | | | |
|---|---|---|---|---|---|---|---|---|
| HBI | InfScore | | Mayo | InfScore | >>> | InfScore | HBI | Mayo |
| 0 | 0.00 | | 0 | 0.00 | merged | 0.00 | 0 | 0 |
| 1 | 0.06 | | 1 | 0.08 | | 0.06 | 1 | |
| 2 | 0.12 | | 2 | 0.17 | | 0.08 | | 1 |
| 3 | 0.18 | | 3 | 0.25 | | 0.12 | 2 | |
| 4 | 0.24 | | 4 | 0.33 | | 0.17 | | 2 |
| 5 | 0.29 | | 5 | 0.42 | | 0.18 | 3 | |
| 6 | 0.35 | | 6 | 0.50 | | 0.24 | 4 | |
| 7 | 0.41 | | 7 | 0.58 | | 0.25 | | 3 |
| 8 | 0.47 | | 8 | 0.67 | | 0.29 | 5 | |
| 9 | 0.53 | | 9 | 0.75 | | 0.33 | | 4 |
| 10 | 0.59 | | 10 | 0.83 | | 0.35 | 6 | |
| 11 | 0.65 | | 11 | 0.92 | | 0.41 | 7 | |
| 12 | 0.71 | | 12 | 1.00 | | 0.42 | | 5 |
| 13 | 0.76 | | | | | 0.47 | 8 | |
| 14 | 0.82 | | | | | 0.50 | | 6 |
| 15 | 0.88 | | | | | 0.53 | 9 | |
| 16 | 0.94 | | | | | 0.58 | | 7 |
| 17 | 1.00 | | | | | 0.59 | 10 | |
| | | | | | | 0.65 | 11 | |
| | | | | | | 0.67 | | 8 |
| | | | | | | 0.71 | 12 | |
| | | | | | | 0.75 | | 9 |
| | | | | | | 0.76 | 13 | |
| | | | | | | 0.82 | 14 | |
| | | | | | | 0.83 | | 10 |
| | | | | | | 0.88 | 15 | |
| | | | | | | 0.92 | | 11 |
| | | | | | | 0.94 | 16 | |
| | | | | | | 1.00 | 17 | 12 |

551 **Table S3: Original and pooled deconvolution cell types.** Originally classified cell types that

552 are ontogenetically related, e.g. Paneth cells, enterocytes, goblet cells and stem cells, were

553 pooled and classified into "epithelial cells" to reduce cell type diversity.

| Original cell type description | Grouped into cell type |
|---|---|
| TA 1 | epis |
| TA 2 | epis |
| Enterocyte Progenitors | epis |
| Immature Goblet | epis |
| Immature Enterocytes 1 | epis |
| Immature Enterocytes 2 | epis |
| Cycling TA | epis |