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Physiologically-Based Pharmacokinetic (PBPK) Model of Morphine and Morphine-3-Glucuronide in Nonalcoholic Steatohepatitis (NASH)

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Abstract

Nonalcoholic steatohepatitis (NASH), the progressive form of nonalcoholic fatty liver disease, is increasing in prevalence. NASH-related alterations in hepatic protein expression (e.g., transporters) and in overall physiology may affect drug exposure by altering drug disposition and elimination. The aim of this study was to build a physiologically-based pharmacokinetic (PBPK) model to predict drug exposure in NASH by incorporating NASH-related changes in hepatic transporters. Morphine and morphine-3-glucuronide (M3G) were used as model compounds. A PBPK model of morphine with permeability-limited hepatic disposition was extended to include M3G disposition and enterohepatic recycling (EHR). The model captured the area under the plasma concentration-time curve (AUC) of morphine and M3G after intravenous morphine administration within 0.82- and 1.94-fold of observed values from three independent clinical studies for healthy adult subjects (n= 6, 10, and 14 individuals). When NASH-related changes in multidrug resistance-associated protein 2 (MRP2) and MRP3 were incorporated into the model, the predicted M3G mean AUC in NASH was 1.34-fold higher compared to healthy subjects, which is slightly lower than the observed value (1.63-fold). Exploratory simulations on other physiological changes occurring in NASH (e.g., moderate decreases in glomerular filtration rate and portal vein blood flow) revealed that the effect of transporter changes was most prominent. Additionally, NASH-related transporter changes resulted in decreased morphine EHR, which could be important for drugs with extensive EHR. This study is an important first step to predict drug disposition in complex diseases such as NASH using PBPK modeling.

Keywords

transporters; drug transport; metabolites; physiologically-based pharmacokinetics; liver

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N.S., S.N. and K.L.R.B. designed the research; N.S. performed the research and N.S., S.N. and K.L.R.B analyzed the data and wrote the manuscript.

Conflict of Interest statement

S.N. is an employee of Certara UK Ltd.. All other authors declared no competing interests for this work.

Introduction

The liver is a major organ of drug elimination. Specific metabolic enzymes and transport proteins in the liver transform and excrete endogenous compounds and xenobiotics, including drugs. Diseases affecting the liver can change the expression and function of these proteins and, therefore, alter drug exposure.^{1, 2} Predicting drug exposure in patients with liver disease is important, especially when developing drugs that are specifically targeted to the liver.

Nonalcoholic fatty liver disease (NAFLD) is the most common cause of chronic liver disease globally.³ NAFLD is characterized by lipid accumulation in hepatocytes and is linked to obesity, insulin resistance and the metabolic syndrome.⁴ It is estimated that up to 25% of the global population suffers from NAFLD.³ The high prevalence of NAFLD is of concern, because its progressive form, nonalcoholic steatohepatitis (NASH), can lead to cirrhosis and hepatocellular carcinoma. Recently, NAFLD has been linked to COVID-19 progression ($p < 0.001$, $n = 202$ patients)⁵, and attention has been called to whether NAFLD patients are more vulnerable to the widespread SARS-CoV-2 infection or if COVID-19 might increase NAFLD progression to NASH.⁶

In NASH, steatosis is accompanied by inflammation, hepatocyte ballooning and fibrosis. Importantly from a pharmacokinetics (PK) and drug safety viewpoint, there is growing evidence that proteins involved in drug absorption, distribution, metabolism and excretion (ADME) are altered in patients with NASH.^{7–12} There is, however, limited information about the functional consequences of NASH-related changes in ADME proteins due to the small number of compounds that have been studied in clinical trials. To date, transporter changes in NASH have been linked to altered PK of acetaminophen glucuronide, morphine glucuronides and the nuclear imaging agent ^{99m}Tc-mebrofenin.^{13–16} Conversely, NAFLD had no effect on apixaban, rosuvastatin or metformin PK^{17, 18}; however, this may not be the case in patients with NASH.

Morphine is an opioid analgesic that is used to treat moderate and severe pain. It undergoes high extraction in the liver and has low oral bioavailability (0.2 – 0.47) due to the extensive first-pass metabolism.^{19–21} Morphine can permeate membranes passively²² and is metabolized by several uridine 5'-diphospho-glucuronosyltransferase (UGT) enzymes. Although it can be metabolized in the intestine,²³ morphine is taken up primarily into the liver by organic cation transporter (OCT) 1, where it is rapidly metabolized mainly by UGT2B7 to its major metabolites morphine-3-glucuronide (M3G) and morphine-6-glucuronide (M6G).^{20, 24, 25} OCT1 activity and UGT2B7 mRNA levels do not appear to be altered in NASH,^{8, 11, 18} which is consistent with similar morphine PK profiles in healthy subjects and patients with NASH.¹⁴ M3G formed in the liver undergoes excretion into bile (canalicular efflux) via multidrug-resistance associated protein (MRP) 2 and basolateral efflux to the systemic circulation via MRP3.^{26, 27} M3G excreted in bile can undergo enterohepatic recycling (EHR).^{20, 21, 27–29} The reported increase in hepatic MRP3 abundance and MRP2 mislocalization in liver tissue from patients with NASH could explain the altered exposure of morphine glucuronides in NASH.^{7, 11, 30}

Physiologically-based pharmacokinetic (PBPK) modeling is a mathematical method that combines drug properties, physiological data and clinical trial information. Clinical observations, e.g. plasma concentration data, can be used to inform model building in a middle-out approach of PBPK modeling. By altering specific physiological parameters (e.g. metabolic enzyme or transporter abundance), PBPK modeling can be utilized to simulate exposure in disease states. It also is useful to examine the contribution of specific changes in system (and drug) parameters on overall PK. This can be beneficial when studying multifactorial diseases such as NASH, where alterations occur not only in ADME proteins, but also in physiological factors, such as body weight.⁴

The aim of this study was to develop a PBPK model to describe the disposition of morphine and M3G in healthy subjects, and extend the model to predict the disposition in patients with NASH based on alterations in transporter abundance and activity. To the best of our knowledge, this is the first PBPK model to include EHR of the metabolite, and luminal metabolism of the metabolite back to parent compound, to evaluate the effect of NASH-related transporter changes on EHR. In addition, simulations were performed to evaluate the contribution of different physiological factors relevant to NASH on morphine and M3G PK.

Methods

PBPK model development for M3G in healthy subjects

An overview of the modeling workflow and a schematic depicting the disposition and enterohepatic recycling of morphine are presented in Figure 1. All simulations were performed using Simcyp simulator software v 18.1 (Certara UK Ltd, Sheffield, UK). A previously published PBPK model of morphine³¹ was used as the basis for model development. A compound file was developed for M3G, using compound specific and measured physico-chemical and blood-binding data (Table 1). The maximum rate of transporter-mediated efflux (V_{max}) for MRP2 and MRP3, as well as the passive permeability clearance in the liver, were optimized using the parameter estimation module with a Maximum Likelihood population-based fitting with M3G serum and urine data from individual healthy subjects.¹⁴ The passive permeability of M3G in the liver was constrained to less than the corresponding permeability of morphine (0.0034 ml/min/million hepatocytes) because glucuronides are expected to exhibit lower passive diffusion compared to the parent.³² A K_p scalar of 0.2 was applied to enable the model to capture the observed peak concentration (C_{max}) of M3G. Only M3G was included in the model because of the large number of data points that were below the lower limit of quantification for M6G, the less abundant but active metabolite of morphine.

An EHR component was included in the model that incorporated MRP2-mediated transport of M3G into bile and subsequent release into the gastrointestinal tract, cleavage of the glucuronide moiety by intestinal (bacterial) β -glucuronidases to release morphine, and subsequent reabsorption of morphine. The effective intestinal permeabilities ($P_{eff,man}$) of morphine ($2.07 * 10^{-4}$ cm/s) and M3G ($0.0266 * 10^{-4}$ cm/s) were predicted based on the polar surface area and number of hydrogen bond donors using the prediction model within the simulator. The relative abundance of hydrolyzing enzymes in the gastrointestinal lumen was set to increase towards the colon, because M3G cleavage/hydrolysis has been shown to

be much higher in the colon than small intestine in laboratory animals.^{28, 33, 34} For example, the initial hydrolysis rate of morphine glucuronide in rat intestines in vitro was up to 178-fold higher in the large intestine compared to the small intestine.³⁴ Therefore, the colon was set as the reference segment and the relative activities were set to 0.0001 in the stomach, 0.001 in the duodenum and jejunum I, 0.1 in the jejunum II and ileum I, and 1.0 in the ileum II – IV compartments and colon to describe this increase. The hydrolysis rate was estimated to be 6000 µl/h/g luminal content, acting on the total concentration in the lumen.^{33, 34} Simulations were performed in the fasted state.

The observed area under the concentration time curve (AUC) values were calculated from the original Ferslew et al.¹⁴ serum data in Phoenix[®] version 8.2. All AUC_{0-∞} values from the simulations also were calculated in Phoenix[®].

Model verification with clinical data from healthy volunteers

Model performance was verified using two independent sets of clinical data describing morphine and M3G disposition after intravenous administration of morphine sulfate.^{21, 35} Plasma concentration-time data were extracted from the publications using the online tool WebPlot Digitizer (<https://automeris.io/WebPlotDigitizer/>). The goodness of fit of the data were evaluated by visual inspection of the simulated plasma concentration time curves with 90% confidence intervals and by comparing simulated plasma AUC values for morphine and M3G, and C_{max} values for M3G, with the observed data. The absolute average fold error (AAFE) was calculated across both healthy and NASH simulations according to equation 1:

$$AAFE = 10^{\frac{1}{n} \left(\sum \left| \log_{10} \left(\frac{Pred}{Obs} \right) \right| \right)} \quad (\text{Equation 1})$$

where Pred and Obs are the predicted and observed values, respectively, of AUC and C_{max}. Predictions of AUC and C_{max} within the two-fold range, i.e. predicted/observed ratios between 0.5- and 2-fold, were regarded as acceptable.

PBPK modeling in NASH incorporating changes in transporters

Simulations were performed using virtual individuals matched for age, weight, height and sex with biopsy-proven noncirrhotic NASH patients.¹⁴ Matching was used because the weight distribution of individuals in the clinical trial was not consistent with either the “Sim-Healthy Volunteers” or the “Sim-Obese” population libraries within the Simcyp Simulator. Previously reported data on MRP2 and MRP3 changes in NASH livers were used to inform the activity changes incorporated into the NASH M3G model. A relative activity factor of 2.5 was used for MRP3 based on immunoblot data.⁷ For MRP2, a relative activity factor of 0.5 was approximated based on information about the mislocalization of MRP2/Mrp2 in NASH,^{7, 30} and the observed ≈ 50% decrease in (1) biliary clearance of the MRP2 substrate ^{99m}Tc-mebrofenin in patients with NASH,¹⁵ and (2) biliary excretion of M3G in rats with methionine-choline diet induced NASH.³⁰ These scaling factors are also in line with recently reported proteomic data showing increased MRP3 and decreased MRP2 abundance.

11

Exploratory simulations

The contribution of different physiological factors that are affected by NASH on morphine and M3G PK was studied by simulation. First, the effect of obesity was investigated by running simulations with the “Sim-Healthy Volunteers”, and the two obesity populations in the default library: “Sim-Obese” (BMI 30–40 kg/m²) and “Sim-Morbidly obese” (BMI >40 kg/m²).³⁶

Further simulations were performed using the “Sim-Obese” population library, but the base glomerular filtration rate (GFR) in the population was set to equal the GFR in the “Sim-Healthy volunteers” population based on published data suggesting that estimated GFR (eGFR) may be comparable in healthy individuals and patients with steatotic livers.^{14, 37} First, the contribution of EHR to morphine and M3G exposure was evaluated by performing simulations where the hydrolysis of M3G to morphine in the gut lumen was disabled in both the control population and the population with NASH-related transporter changes. Next, the effects of NASH-related alterations in renal function and portal vein blood flow were examined. In contrast to data reported for obese patients,³⁶ patients with NASH may have slightly decreased renal clearance (14% lower eGFR than healthy controls) and are at a higher risk of chronic kidney disease.³⁸ Impaired GFR was incorporated into the model by increasing the serum creatinine values of the obese population to obtain approximately 14% lower mean GFR in the virtual obese subjects than in the virtual healthy population. GFR was calculated using the Modification of Diet in Renal Disease (MDRD) method. The effects of portal vein blood flow were studied in simulations where portal vein blood flow (described as % of cardiac blood flow) was reduced by 15% and 30% to mimic observations in patients with NASH and different stages of fibrosis.³⁹ Finally, the effect of altered MRP2 and MRP3 transport on morphine and M3G was simulated, along with simulating the combined effects of the other physiological changes. Ten trials of ten subjects, 50% female, age 33–63 years were simulated in each of the exploratory simulations.

Results

Model development and verification

Model development was performed using the M3G serum and urine data from individual healthy volunteers (n = 14).¹⁴ The parameter values used in the final model for M3G are compiled in Table 1. Based on visual inspection of the plasma concentration-time profiles, the model captured the M3G data adequately for the dataset used for model development as well as the two datasets used for model verification^{21, 35} (Figure 2). The ratios of the predicted to observed mean AUC and C_{max} values were within the two-fold acceptance criterion and ranged from 0.82–1.49, except the AUC_{0–∞} ratio for morphine in the model development set which was 1.94 (Table 2).

PBPK modeling in NASH incorporating changes in transporters

The plasma PK of morphine and M3G were simulated by incorporating NASH-related transporter changes into the developed model. The performance of the model was similar to the simulations for healthy subjects, with the ratio of predicted to observed values ranging from 0.81–1.87 (Table 2, Figure 2), which met the two-fold acceptance criterion. Across all

simulations, the AAFE of morphine AUC was 1.48 and the AAFEs of M3G AUC and C_{\max} were 1.12 and 1.13, respectively. M3G exposure (AUC_{0-8h}) was predicted to be 1.34-fold higher in patients with NASH compared to healthy controls whereas the observed value was 1.63-fold. Similarly, the predicted M3G C_{\max} change in patients with NASH (1.36-fold) was lower than the observed value (1.54-fold).

Evaluation of individual simulations

In addition to evaluating the mean fits, observed data for each subject from Ferslew et al.¹⁴ were overlaid with the plasma concentration-time profiles from the ten matched virtual individuals for both healthy subjects and patients with NASH. Characteristics of virtual individuals with aberrant simulated profiles were investigated to uncover possible reasons for these abnormalities. Examples of cases are presented in Figure 3. High OCT1 values (> 2.8-fold of average) resulted in lower morphine concentrations in some simulated subjects (Subject A2 and B2), whereas a low OCT1 value (0.11-fold of control) increased morphine concentrations (Subject B1). In these examples, individuals with a low MRP3/MRP2 ratio (0.15) and high OCT1 activity (> 2.1-fold of average) showed relatively low and sustained concentrations of M3G over the simulation period (Subject A1 and B2). M3G concentrations were also lower with a combination of lower than average UGT2B7 activity (0.78-fold of average) and a low MRP3/MRP2 ratio (0.12) (Subject B3). In this individual, OCT1 activity was slightly below average (0.75-fold of average) and the slope of the M3G terminal phase was similar to other curves despite the low MRP3/MRP2 ratio. For Subject A2, a combination of high OCT1 (3.05-fold of average), high metabolism to M3G (1.40-fold of average) and a high MRP3/MRP2 ratio (7.30) resulted in high M3G exposure.

Exploratory simulations

There are differences in the demographics of the populations simulated using the Simcyp library populations for obese and morbidly obese subjects (Supplementary data Table S1). Body and liver weight were increased as was the volume of distribution at steady-state (V_{ss}). Notably, the GFR in simulated obese and morbidly obese individuals was 30% and 65% higher, respectively, than healthy volunteers. Hepatic blood flow was not different between the populations. Morphine AUC_{0-12h} was not markedly affected by obesity (<20% decrease), but mean M3G AUC_{0-12h} decreased by 25% in obese and 45% in morbidly obese populations compared to the healthy population (Figure 4A and B). In subsequent simulations, the obese population was used, but the GFR was set equal to that of the healthy population.

The contribution of EHR was simulated by disabling the hydrolysis of M3G to morphine in the gastrointestinal lumen, which is the major component of morphine EHR.^{28, 29} The M3G AUC_{0-12h} was 21% lower when the hydrolysis of M3G to morphine was disabled in the model; the morphine AUC_{0-12h} was decreased by 9% (Figure 4 C and D, Table 3). When NASH-related transporter changes were included in the model, the difference in M3G AUC_{0-12h} with and without EHR was approximately 13%.

To examine the effects of reduced GFR, the baseline healthy GFR was modified to simulate the 14% decrease observed in patients with NASH by Targher et al.³⁸ The reduction in GFR

had little effect on morphine exposure, but increased M3G exposure by 15% (Table 3, Supplementary Figure S1 A and B). Reducing portal vein blood flow by 15% and 30% in simulations led, in both cases, to a 6% decrease in M3G exposure, but the 30% reduction in portal vein flow increased morphine AUC_{0–12h} by 17% (Table 3, Supplementary Figure S1 C and D).

Of the tested NASH-related physiological changes, NASH-mediated transporter alterations (i.e., simultaneous scaling of MRP2 and MRP3 relative activity by 0.5- and 2.5-fold of control, respectively) had the highest effect on M3G exposure (Table 3, Supplementary Figure S1 E and F). M3G AUC was increased 40%, but morphine exposure was not markedly altered. The combination of transporter changes, reduced GFR and a 30% decrease in portal vein blood flow increased the exposure of M3G by 54% compared to the control group (Table 3, Supplementary Figure S1 G and H).

Discussion

Alongside the obesity epidemic, the prevalence of NAFLD and NASH is increasing rapidly.³ Recently, the worldwide COVID-19 pandemic has raised concerns that NAFLD may predispose patients to SARS-CoV-2 infection and COVID-19 complications, and that COVID-19 could expedite NAFLD progression.⁶ However, this remains to be studied. At present, NASH is one of the top three causes of hepatic transplants in the U.S. and there is no direct treatment for NASH beyond lifestyle changes.³ NASH patients are often burdened by other diseases such as cardiovascular disease and type 2 diabetes that require drug treatment.^{3, 4} Therefore, understanding NASH-associated alterations in drug PK is not only important for the development of drugs to treat NASH, but also for ensuring safe treatment of other comorbidities in patients with NASH using currently-marketed drugs.

One of the challenges of predicting exposure in this patient population is that NAFLD is progressive. Liver size and composition can vary and it has been shown that there are differences in transporter abundance not only between normal and NAFLD livers, but also between simple steatosis and NASH, as well as NASH with fatty liver and NASH without fatty liver.^{7, 11} However, it is difficult to determine the stage of liver disease of a patient or a study subject, because the most reliable way to evaluate steatosis and fibrosis in NAFLD and to diagnose NASH is liver biopsy, which is invasive.⁴ Typically, the diagnosis is determined by the presence of hepatic steatosis based on imaging and histology, no evidence of significant alcohol consumption and no competing causes of steatosis or chronic liver disease.

Due to the challenges of diagnosing NASH stage and evaluating the associated degree of changes in transporters and metabolic enzymes, it may be difficult to make PK predictions on an individual level. However, PBPK modeling can be used to gauge interindividual variability and the risk of altered exposure and potential adverse effects. Furthermore, it can be utilized to study the effects of specific factors that cannot be distinguished from *in vivo* data either due to restrictive inclusion criteria in clinical studies or the complexity of the *in vivo* system.

In this study, a PBPK model of morphine and M3G was constructed and verified using data from healthy volunteers. By incorporating NASH-related changes in transporter abundance and activity into the model, M3G exposure was predicted to increase 1.34-fold compared to healthy subjects, consistent with increased M3G exposure in patients with NASH.¹⁴ The increase was driven by the combined increased MRP3 (2.5-fold) and decreased MRP2 function (0.5-fold). The changes in MRP2 and MRP3 function had the greatest effect on M3G exposure in our exploratory simulations. Although there is some evidence for increased total protein abundance of MRP2 in NASH, changes in glycosylation and localization in hepatocytes suggest that decreased abundance at the canalicular membrane contribute to decreased function.^{7, 9} A decrease in MRP2 function is supported by a 48% decrease in biliary AUC of M3G observed in a rat model of NASH, which has similar histologic hepatic hallmarks as human NASH.³⁰ This decrease also could explain the observed 1.6-fold increase in hepatic ^{99m}Tc-mebrofenin exposure and the ~2-fold decrease in biliary clearance observed in patients with NASH.¹⁵

Obesity is highly associated with NAFLD and NASH and it is estimated that the majority of severely obese patients undergoing bariatric surgery have NAFLD.⁴⁰ Therefore, simulations were performed using obese and morbidly obese virtual populations.³⁶ A major difference in these virtual populations compared to healthy volunteers was the increase in GFR (Supplementary Table S1), which is propagated as increased renal clearance. This increase in renal clearance explains the simulated decrease in morphine and M3G exposure in the obese and morbidly obese populations, because renal excretion is the primary route of elimination for morphine glucuronides and excretion is related to renal function.^{21, 41} The simulated decrease in M3G exposure in the obese populations is contradictory to a study by de Hoogd et al.⁴², where the plasma concentration of M3G was higher in morbidly obese patients than in healthy subjects. However, NAFLD/NASH status was not evaluated by these authors, so the effect of obesity cannot be distinguished from possible NASH-related transporter alterations.

NASH is associated with an increased risk of chronic kidney disease (CKD) and eGFR is reported to be slightly decreased (8–14%) in overweight subjects with NASH (n=80) compared with nonsteatotic matched controls (n=80), and in morbidly obese NASH subjects (n=37) compared with subjects with simple steatosis (n=111).^{37, 38} Although CKD risk also was reported to be increased in a Japanese overweight population with NASH, statistical testing for differences in eGFR were inconclusive.⁴³ For the simulations corresponding to the patients with NASH in the Ferslew et al.¹⁴ study, GFR was not adjusted because there was no significant difference in eGFR between the study cohorts. However, in the present study, the effect of decreased GFR was simulated in the exploratory simulations and a 14% decrease in GFR resulted in a 15% increase in M3G exposure. The combination of reduced GFR with NASH-related transporter changes increased M3G exposure to similar levels as observed in the clinical study by Ferslew et al.¹⁴. Decreases in portal vein blood flow primarily affected morphine exposure, which is consistent with previous work showing that morphine clearance is sensitive to changes in cardiac output in both adults and children.³¹

One of the novel features of this PBPK model is the incorporation of morphine EHR. In the exploratory simulations, EHR was shown to account for 21% of M3G exposure (M3G

AUC_{0–12h} decreased from 628 to 494 nM·h in the absence of M3G hydrolysis). Due to the extensive first-pass metabolism of morphine to M3G, the effect on morphine exposure was slightly less (9%). The effect of EHR was most pronounced when the ratio of MRP3 to MRP2 transport was low (i.e., canalicular efflux was dominant) and OCT1 transport was higher than average, resulting in sustained M3G concentration profiles (Figure 3). In line with this, EHR had less of an impact in the NASH population, because M3G excretion into bile via MRP2 is diminished in NASH (Table 3, Figure 4.). This difference in EHR between healthy subjects and patients with NASH could be important for some drugs (e.g., ezetimibe) that have an intestinal target and undergo extensive glucuronidation and EHR.⁴⁴ Another factor to consider regarding EHR is possible alterations in the intestinal microbiome in NASH that could affect glucuronide cleavage by gut bacteria. These alterations are supported by observed increases in secondary bile acids that are formed in the gut by the gut microflora in patients with NASH.⁴⁵

A limitation of this study is that it is focused on M3G, which is the inactive glucuronide metabolite of morphine. Only M3G data were incorporated in the model due to the large number of M6G concentrations that were below the limit of quantification and the resulting low confidence in estimating the clearance values for MRP2 and MRP3 efflux. Based on the clinical data, however, NASH effects appear similar for M6G and M3G.¹⁴

To avoid overparameterization, the model focused on the major transporters for morphine (OCT1) and M3G (MRP2 and MRP3). However, *in vitro* data indicate that morphine also is a substrate of OATP2B1,⁴⁶ and clinical data suggest that P-glycoprotein may affect morphine absorption.⁴⁷ Therefore, inclusion of these transporters could be important for modeling oral dosing data, which was not the focus in this study. In addition, studies in HEK293-OCT2 cells suggest that morphine is a substrate of the renal transporter OCT2.⁴⁸ However, the influence of OCT2 on morphine disposition is expected to be low since only < 10% of morphine is recovered unchanged in urine.^{20, 21} Although M3G is not a substrate of OATP2B1,⁴⁸ data from uptake studies in hepatocytes suggest that OATP1B transporters are involved in M3G uptake into the liver.¹¹ The significance of OATP1B transporters, which are altered in NASH,^{9, 11, 12, 15} on M3G disposition in healthy subjects and in NASH remains to be further evaluated to address identifiability issues between OATP1B uptake and MRP3 efflux at the hepatic basolateral membrane. Even a clinical study focusing on the effects of OATP1B1 genotypes on M3G disposition could only help to answer this question if the OATP1B contribution is significantly affecting the PK of M3G.

Another point to consider is that although NAFLD and NASH affect the liver directly, the disease could result in compensatory changes elsewhere in the body. Unfortunately, there is a lack of information on transporter alterations in the intestine and kidneys of NASH patients, but changes might be expected based on findings from rodent models of NASH.^{49, 50} Active secretion in the kidney does not appear to be significant for M3G, but renal metabolism of morphine does contribute to M3G formation.^{20, 41, 51} More clinical data are needed to incorporate NASH-related changes in tissues other than the liver, for example the intestine, into the PBPK model. Another limitation is that in the model, body surface area is used to scale organ size, which results in larger organs for obese subjects. However, it is known that in addition to increased liver size with obesity, the composition of the liver is

altered in NASH by definition, and the functional mass might be decreased with disease progression. Further work and clinical data are required to study the effects of these changes on drug disposition.

Conclusions

The PBPK model constructed here for morphine and M3G highlighted the importance of NASH-related transporter changes for M3G disposition, and was able to capture some of the changes in M3G exposure observed in patients with NASH. Other physiological factors that are altered in NASH had less pronounced effects on M3G simulations. They should, however, be accounted for within a PBPK model for NASH since these minor effects can become more relevant in combination with each other. The contribution of EHR to M3G exposure was decreased in NASH due to altered M3G transport. More clinical data are needed to better understand NASH-related changes in the liver and other tissues in order to incorporate these into PBPK modeling. This work is an important first step, because modeling is an indispensable tool to study and predict drug disposition in complex diseases such as NASH. However, since the physiological changes in NASH are so complex, we focused on the evaluation and relevance of a limited number of transporters for the present study. More clinical studies that consider the fate of the metabolite are needed to extend our understanding of NASH, so that future models can be used to draw conclusions about the effects of NASH and its progression on other drugs/metabolites.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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Study Highlights

- What is the current knowledge on the topic?
Nonalcoholic steatohepatitis (NASH) is increasing in prevalence. NASH-related alterations in physiology and protein expression (e.g., transporters) may affect drug exposure by altering drug disposition and elimination.
- What question did this study address?
This study investigated the use of physiologically-based pharmacokinetic (PBPK) modeling to predict drug exposure in NASH. NASH-related changes in hepatic transporters were incorporated and the effects of other NASH-related physiological changes were investigated using morphine and morphine-3-glucuronide (M3G) as model compounds.
- What this study adds to our knowledge?
Simulations revealed that NASH-related changes in hepatic transporters can explain a majority of the increased M3G exposure in NASH patients. The effect of transporters was more prominent than moderate decreases in GFR and portal blood flow. In addition, transporter changes resulted in decreased morphine enterohepatic recycling.
- How might this change drug discovery, development, and/or therapeutics?
PBPK models incorporating NASH-related changes may aid in identifying drugs with increased risk of altered exposure, understanding sources of interindividual variability and helping ensure safe and efficacious drug therapy in this patient population.

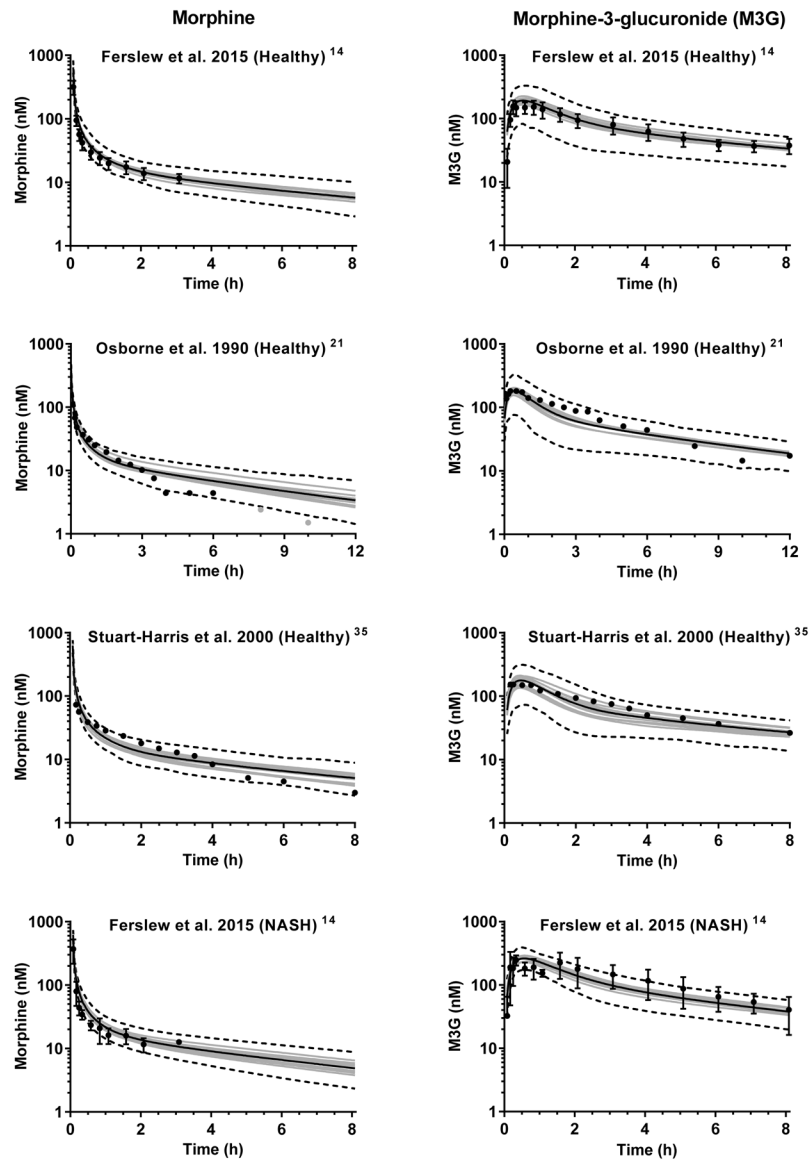


Figure 2.

Simulated plasma concentration-time profiles of morphine (left panel) and morphine-3-glucuronide (M3G; right panel) after intravenous administration of 3.76 mg morphine (corresponding to 5 mg morphine sulfate) to virtual subjects in ten trials. The number of subjects in each virtual trial was matched with the corresponding clinical study. Gray lines represent the mean of one trial and the black solid line represents the mean of all trials. Dashed lines show the simulated 5th and 95th percentiles. Simulations are overlaid with mean plasma/serum data for fourteen¹⁴, ten²¹ and six³⁵ healthy volunteers and seven patients with nonalcoholic steatohepatitis (NASH).¹⁴ The error bars in the observed data from Ferslew et al.¹⁴ show the 95% confidence interval. The light gray points in the Osborne et al.²¹ morphine panel indicate observed values close to the analytical lower limit of quantification.

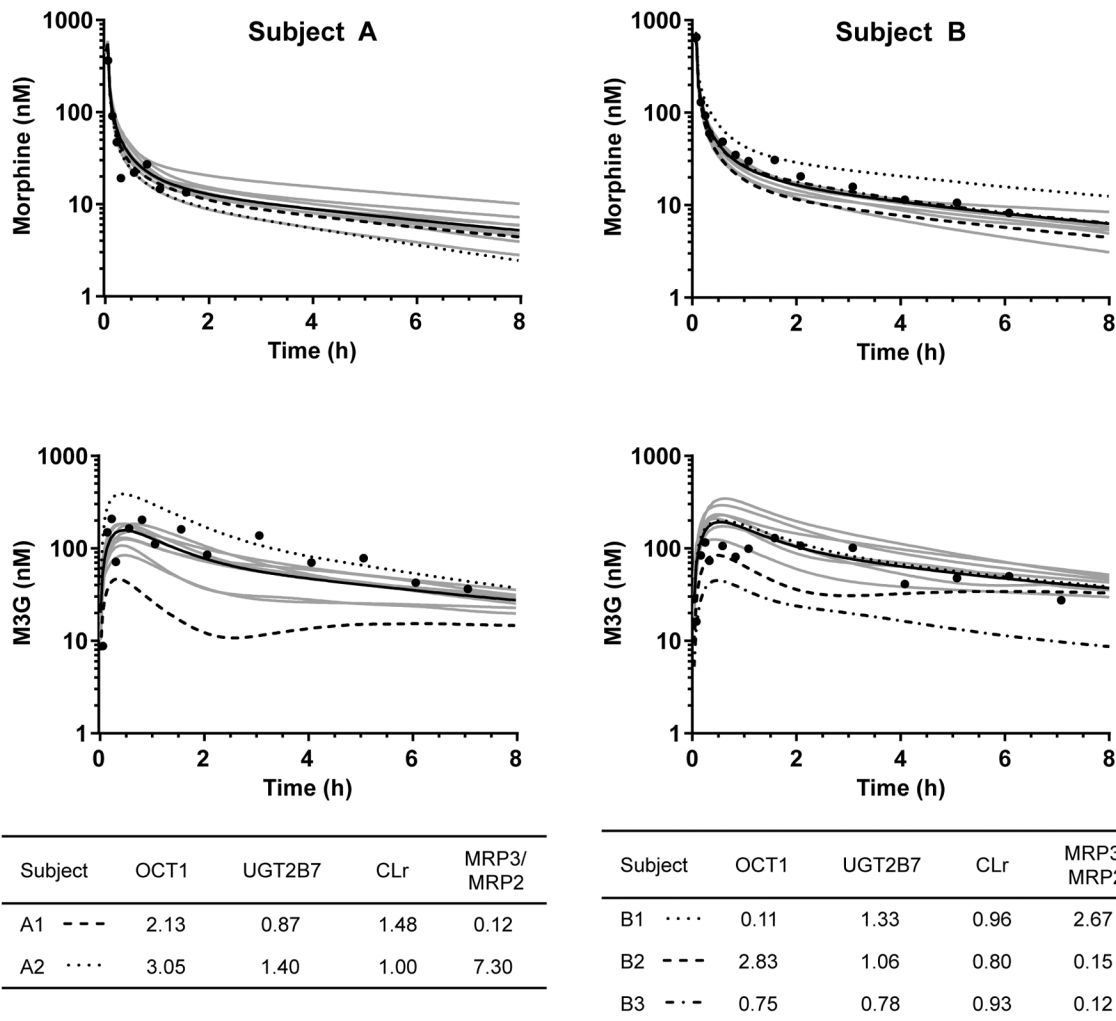


Figure 3.

Two examples of simulated individual plasma concentration-time profiles of morphine and morphine-3-glucuronide (M3G) in healthy subjects. Shown are observed concentrations (black circles), the mean of simulated profiles of ten virtual subjects matched for age, weight, height and sex with the corresponding study subject (black line) and individual profiles of the virtual subjects. With this matching, virtual individuals share characteristics linked to body surface area and age (e.g., liver weight), but have individual enzyme and transporter phenotypes. Specific features of highlighted individuals are given in tables. Values in the OCT1, UGT2B7 and CL_r columns are normalized to the population mean. OCT1, maximal hepatic uptake clearance of morphine; UGT2B7, maximal hepatic clearance of morphine to M3G; CL_r, renal clearance of morphine; MRP3/MRP2, ratio of maximal basolateral and canalicular efflux clearance.

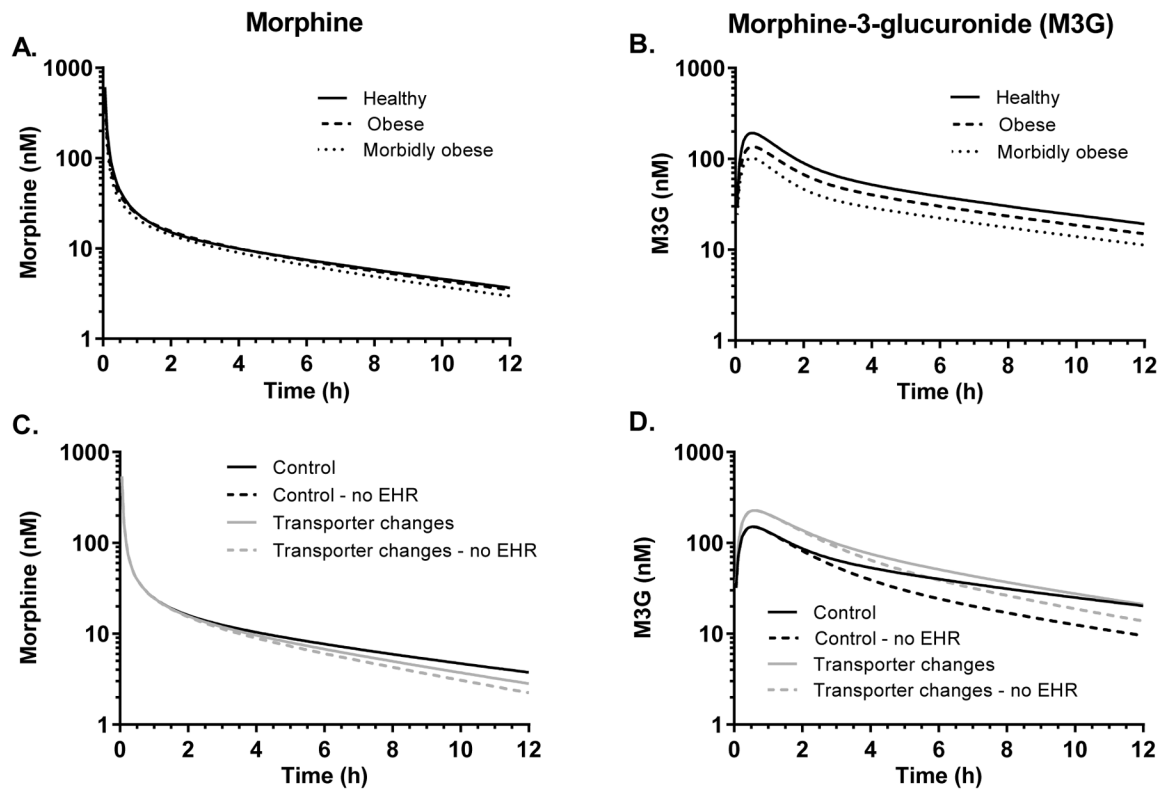


Figure 4.

Simulated plasma concentration-time profiles of morphine (left panel) and morphine-3-glucuronide (M3G; right panel) from exploratory simulations. One hundred virtual individuals were simulated using the Simcyp library healthy volunteers, obese and morbidly obese populations (A and B). In panels C and D, simulations were performed using the “Sim-obese” population where glomerular filtration rate (GFR) was set to match the healthy volunteers (Control) and the same populations with NASH-related transporter changes with and without enterohepatic recycling (EHR). In the simulations with no EHR, the luminal hydrolysis of M3G to morphine was disabled. All results are shown as the mean of all simulations. In C, the “Control – no EHR” and the “Transporter changes – no EHR” profiles are overlapping.

Table 1.

Final input parameters used in the morphine-3-glucuronide model.

Parameter	Value	Reference
Molecular weight	461.46 g/mol	
LogP	-1.1	52
Compound type	Ampholyte	
pKa1	2.86 (acid)	52
pKa2	7.93 (base)	53
Blood-to-plasma ratio	0.61	54
Fraction unbound in plasma	0.85	41
Main plasma binding component	Albumin	
Distribution	Full PBPK model	
V _{ss}	Predicted	Method 2 ⁵⁵
K _p scalar	0.2	Optimized
Renal clearance	8.4 l/h (for 70 kg subject)	20
Permeability limited liver		
Passive permeability	0.0006 ml/min/10 ⁶ hepatocytes	Optimized
MRP3		
V _{max}	7300 pmol/min/10 ⁶ hepatocytes	Optimized
K _m	500 μM	26
MRP2		
V _{max}	590 pmol/min/10 ⁶ hepatocytes	Optimized
K _m	50 μM	27
Polar surface area	149 Å ²	Pubchem
Number of hydrogen bond donors	5	Pubchem
P _{eff,man}	0.0266*10 ⁻⁴ cm/s	Predicted
f _{gut}	1	Assumed
Gut luminal hydrolysis CL _{int}	6000 μl/h/g luminal contents	33, 34

CL_{int}, intrinsic clearance; f_{gut}, fraction of drug unbound in the enterocyte; LogP, log of the octanol-water partition coefficient for the neutral compound; K_m, Michaelis-Menten constant; K_p, equilibrium distribution ratio; MRP, multidrug-resistance associated protein; P_{eff,man}, effective human jejunum permeability; V_{max}, maximum rate of transporter-mediated efflux; V_{ss}, volume of distribution at steady-state.

Table 2.Predicted and observed AUC of morphine and AUC and C_{max} of M3G in plasma/serum.

	Predicted	Observed	Ratio predicted/observed
Model development			
<i>Ferslew et al. 2015</i> (n=14) ¹⁴			
Morphine AUC _{0-∞} (nM*h)	211 ± 15 [190 – 235]	109 ± 45	1.94 [1.74 – 2.16]
Morphine AUC _{0-last} (nM*h) ^a	115 ± 4 [109 – 121]	88 ± 37	1.31 [1.24 – 1.38]
M3G AUC _{0-8h} (nM*h)	613 ± 59 [541 – 737]	588 ± 175	1.04 [0.92 – 1.25]
M3G C _{max} (nM)	193 ± 18 [165 – 226]	211 ± 62	0.91 [0.78 – 1.07]
Model verification			
<i>Osborne et al. 1990</i> ^b (n=10) ²¹			
Morphine AUC _{0-∞} (nM*h)	170 ± 24 [144 – 223]	114 ± 18	1.49 [1.26 – 1.96]
M3G AUC _{0-12h} (nM*h)	639 ± 55 [542 – 712]	784 ± 110	0.82 [0.69 – 0.91]
M3G C _{max} (nM)	182 ± 15 [155 – 204]	204 ± 50	0.89 [0.76 – 1.00]
<i>Stuart-Harris et al. 2000</i> ^b (n=6) ³⁴			
Morphine AUC _{0-8h} (nM*h)	149 ± 12 [129 – 165]	135 ± 31	1.10 [0.96 – 1.22]
M3G AUC _{0-8h} (nM*h)	499 ± 84 [378 – 638]	511 ± 128	0.98 [0.74 – 1.25]
M3G C _{max} (nM)	178 ± 28 [135 – 211]	167 ± 38	1.07 [0.81 – 1.26]
NASH simulation			
<i>Ferslew et al. 2015</i> (n=7) ¹⁴			
Morphine AUC _{0-∞} (nM*h)	183 ± 20 [160 – 219]	98 ± 36	1.87 [1.63 – 2.23]
Morphine AUC _{0-last} (nM*h) ^a	104 ± 6 [96 – 115]	77 ± 30	1.35 [1.25 – 1.49]
M3G AUC _{0-8h} (nM*h)	823 ± 67 [699 – 920]	961 ± 331	0.86 [0.73 – 0.96]
M3G C _{max} (nM)	263 ± 23 [231 – 292]	325 ± 72	0.81 [0.71 – 0.90]
NASH effect ^c			
Morphine AUC _{0-last}	0.90	0.88	1.02
M3G AUC _{0-8h}	1.34	1.63	0.82
M3G C _{max}	1.36	1.54	0.88

Data are presented as the mean ± SD of ten simulated trials with the corresponding number of subjects (Sim-Healthy Volunteers) as the clinical trial in each simulated trial. For the NASH simulations, NASH-related transporter changes were incorporated into the model. The range of means from individual simulated trials are shown in brackets.

^aThe area under plasma concentration-time curve (AUC) range from simulations was matched to the available data from the matching clinical study subjects.

^bReported values were transformed to correspond to a dose of 5 mg morphine sulfate.

^cThe NASH effect was calculated as the ratio of the parameter in the NASH simulation and the model development simulation. C_{\max} , peak concentration in plasma.

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Table 3.

Results from the simulations investigating the effect of enterohepatic recycling (EHR) and different nonalcoholic steatohepatitis (NASH)-related physiological changes on morphine and morphine-3-glucuronide (M3G) AUC.

Condition	Morphine		M3G	
	AUC _{0-12h} (nM*h)	Ratio (condition/ control) ^a	AUC _{0-12h} (nM*h)	Ratio (condition/ control) ^a
Control (Obese with healthy GFR)	175 ± 7		628 ± 88	
No EHR	160 ± 5	0.91 ± 0.05	494 ± 76	0.79 ± 0.16
Transporter changes with no EHR	160 ± 5	0.91 ± 0.05 (1 ± 0.05) ^b	781 ± 84	1.24 ± 0.22 (1.58 ± 0.30) ^b
Reduced GFR by 14%	178 ± 7	1.02 ± 0.06	722 ± 102	1.15 ± 0.23
Reduced portal vein blood flow by 15%	188 ± 6	1.07 ± 0.06	612 ± 85	0.97 ± 0.19
Reduced portal vein blood flow by 30%	205 ± 6	1.17 ± 0.06	591 ± 82	0.94 ± 0.19
Transporter changes	166 ± 6	0.95 ± 0.05 (1.04 ± 0.05) ^c	879 ± 86	1.40 ± 0.24 (1.13 ± 0.16) ^c
Transporter changes and reduced GFR by 14%	169 ± 7	0.97 ± 0.05	1022 ± 100	1.63 ± 0.28
Transporter changes, reduced GFR by 14% and portal vein blood flow by 30%	195 ± 6	1.11 ± 0.06	970 ± 95	1.54 ± 0.26

One hundred obese fasted subjects, 50% female, age 33–63 years were simulated for each condition. The control represents the virtual obese population with the glomerular filtration rate (GFR) equal to healthy volunteers. Transporter changes include simultaneous scaling of multidrug-resistance associated protein (MRP) 2 and MRP3 relative activity by 0.5- and 2.5-fold of control, respectively. In the simulations with no EHR, the hydrolysis of M3G in the gut lumen to morphine was disabled. Data are reported as mean ± SD area under the plasma concentration-time curve from 0 to 12 h (AUC_{0-12h}).

^aThe ratio (condition/control) represents the simulated condition to the control value (Obese with healthy GFR), except where otherwise noted.

^brepresents a comparison between the ‘Transporter changes with no EHR’ and ‘No EHR’;

^crepresents a comparison between ‘Transporter changes’ and the ‘Transporter changes with no EHR’.