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Features of an altered AMPK metabolic pathway in Gilbert's Syndrome, and its role in metabolic health

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Energy metabolism, involving the ATP-dependent AMPK-PgC-Ppar pathway impacts metabolic health immensely, in that its impairment can lead to obesity, giving rise to disease. Based on observations that individuals with Gilbert's syndrome (GS; *UGT1A1**28 promoter mutation) are generally lighter, leaner and healthier than controls, specific inter-group differences in the AMPK pathway regulation were explored. Therefore, a case-control study involving 120 fasted, healthy, age- and gender matched subjects with/without GS, was conducted. By utilising intra-cellular flow cytometry (next to assessing *AMPK α 1* gene expression), levels of functioning proteins (phospho-AMPK α 1/ α 2, PgC 1 α , Ppar α and γ) were measured in PBMCs (peripheral blood mononucleated cells). In GS individuals, rates of phospho-AMPK α 1/ α 2, -Ppar α / γ and of PgC 1 α were significantly higher, attesting to a boosted fasting response in this condition. In line with this finding, *AMPK α 1* gene expression was equal between the groups, possibly stressing the post-translational importance of boosted fasting effects in GS. In reflection of an apparently improved health status, GS individuals had significantly lower BMI, glucose, insulin, C-peptide and triglyceride levels. Herewith, we propose a new theory to explain why individuals having GS are leaner and healthier, and are therefore less likely to contract metabolic diseases or die prematurely thereof.

The metabolism significantly impacts energy turnover on a cellular and systemic level. Under physiological conditions, energetic homeostasis is warranted by metabolising macro-nutrients (ATP refuelling) on the one hand, and by adaptive measures to compensate for energetic surplus on the other. In situations of chronic pathophysiological deregulation of key factors therein, metabolic complications including obesity can result. Secondary diseases are promoted, including type II diabetes mellitus (DM II), a condition that is among the most prominent pathological consequences of energetic misbalance^{1,2}.

Healthy carriers of the *UGT1A1**28 promoter mutation, that is characteristic for the benign condition of Gilbert's syndrome (GS; *i.e.* M. Meulengracht), present with moderate unconjugated hyper-bilirubinaemia. The underlying polymorphism is characterised by an additional -TA repeat in the TATA-sequence of the *UGT1A1* promoter, to yield (TA)₇/(TA)₇ instead of (TA)₆/(TA)₆³. This missense mutation results in a reduced *UGT1A1* enzyme function, leading to a decreased conjugation of bilirubin. Next to higher levels of unconjugated bilirubin (UCB) and in the absence of any other adverse symptoms, significantly lower body mass indices (BMI)⁴, improved glucose and lipid profiles⁵, and a resulting lower prevalence of DM II and of other chronic metabolic/inflammatory disorders⁶⁻⁸ have been reported specifically for this group. This imposed the assumption that crucial energetic switches such as AMPK (heterotrimeric AMP-activated ser/thr kinase), are likely positively affected

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Variables	Males		<i>p</i> -value	Females		<i>p</i> -value
	GS	C		GS	C	
Subjects [n]	40	40	1.000	20	20	1.000
Median age [yrs]^	30 (19)	31 (20)	0.965	40 (20)	40 (18)	0.766
Subjects aged < / ≥ 35 yrs [n/n]	24/16	24/16	1.000	9/11	9/11	1.000
Age of subjects < / ≥ 35 yrs^	26.5 (6.0)/45.5 (15.0)	27.0 (6.0)/45.5 (14.0)		29 (5)/48 (19)	29 (5)/46 (14)	
UCB concentration [μM] ^o	35.3 (±10.4)	9.7 (±3.5)	0.000*	28.9 (±6.8)	8.4 (±3.1)	0.000*
UGT1A1*28 genotype/-TA repeats [% 7_7/6_7/6_6] [□]	86.8/10.5/2.5	5.3/50/44.7	0.000*	94.7/5.3/0	5.3/45/45	0.000*
Health food consumption [times/week] ^o	28.6 (±9.3)	24.5 (±13.2)	0.153	32.0 (±8.0)	31.0 (±7.0)	0.611
Snack food consumption [times/week]^	10.5 (9.3)	10.0 (7.0)	0.941	9.0 (11.5)	13 (11.5)	0.876
Red meat consumption [times/week]^	3.0 (6.5)	3.0 (6.0)	0.891	0 (3)	3 (4)	0.103
Alcoholic drinks consumption [times/week]^	1 (3)	0 (2)	0.419	0 (4)	1 (3)	0.945
Overall activity [times/week]^	5.0 (3.5)	6.5 (5.5)	0.025*	6.0 (3.5)	4.0 (4.1)	0.069 ^T
Endurance exercise [times/week]^	2 (2)	3 (5)	0.003*	4.0 (3.0)	2.0 (2.2)	0.074 ^T
Resistance exercise [times/week]^	1 (2)	1 (2)	0.908	0.3 (2.0)	0.0 (1.8)	0.667

Table 1. Demographical and behavioural description of the BiliHealth study population (all subjects).

(Table 1 provides a comparative (GS versus C) demographical and behavioural description of all subjects of the BiliHealth study. Based on data distribution, means^o or medians[^] are presented. For parametric data mean ± sd, for non-parametric distribution, medians and IQR (inter-quartile range) are given. P-values of ≤ 0.05* indicate significant differences; trends are reflected by $p \leq 0.1$ ^T. [□]Insertion of additional -TA repeats in the UGT1A1*28 promoter region; 7_7: Gilbert's syndrome, 6_7: heterozygous individuals, 6_6: wild type. Abbreviations UCB: unconjugated bilirubin; UGT1A1-genotype: UDP glucuronosyltransferase 1A1 genotype.

in the condition of GS. Representing one of the most important energetic controllers and bottleneck of all energy consuming cellular processes, AMPK α 1/ α 2 catalytic activity together with subsequent downstream metabolic effectors (Pgc 1 α , peroxisome proliferator-activated receptor gamma coactivator 1- α ; Ppar α and γ , peroxisome proliferator-activated receptors α and γ ; Sirt-1, sirtuin-1; FGF-21, fibroblast growth factor 21) were explored in terms of inter-group (GS and controls) differences in (activated) protein levels.

The core regulatory unit studied, AMPK, is a member of a metabolite sensing protein kinase family, that is present in all eukaryotes⁹, and retained in all cell types for regulating energy turnover. It is allosterically activated by increasing levels of ADP and AMP, and therefore considered primarily as a “fuelling-gauge” recognizing ATP depletion (as in fasting), limiting further energy consumption¹⁰. Together with a decline in ATP, upstream kinase activity determine AMPK's activity through its phosphorylation status^{11,12}. Active AMPK subsequently inactivates enzymes responsible for cholesterol-, fatty acid synthesis and gluconeogenesis. For years, this very mechanism has been exploited to routinely treat DM II, by using the anti-diabetic drug Metformin to increase AMPK phosphorylation, and ultimately improve glucose metabolism^{13,14}.

Another important effect of AMPK activation includes the post-translational phosphorylation of Pgc 1 α , which is a positive regulator of energy consuming events such as oxidative processes (including mitogenesis and browning of adipose tissue), and adaptive thermogenesis. Its activity is furthermore fuelled in conditions of physical stress¹⁵, and is enhanced by the enzyme Sirt-1, another determinant of energy homeostasis¹⁶.

In immediate response to active Pgc 1 α and Sirt-1¹⁶, Ppars, an isotypic group of three (α , β /8 and γ ¹⁷) nuclear receptor phospho-proteins¹⁸ and transcription factors¹⁹, are expressed. Means of their activation include phosphorylation through AMPK and ligand binding including fatty acids²⁰. Ppars are specifically abundant in certain tissues including the liver, brain, muscle and cells of the immune system²¹. They occur ubiquitously in all cells²², as they control the expression of genes involved in adipogenesis and lipid metabolism. Therefore, Ppars are considered as crucial networkers of energy- and nutrient-catabolism^{23–25}, which is why they are strongly implicated in the development and treatment of the metabolic syndrome^{26,27}. Further associated with metabolic regulations, and upon Ppar α signaling, expression of FGF-21 takes place in the liver and adipose tissue²⁸, from where it reaches the circulatory system. In animal studies, administration of this factor has been shown to positively impact metabolism in obese mice²⁹ and diabetic monkeys³⁰. Furthermore, serum FGF-21 has been suggested as a potential cardio-metabolic biomarker for humans²⁸.

So far, there is no literature available on a possible role for the UGT1A1*28 polymorphism and/or elevated levels of UCB in these complex regulations. As yet, only one recent study³¹ points out a connection between circulating bilirubin (BR) and Ppar α , using an *in vitro* approach. This very report expands the field of known physiological bilirubin functions and activities, including antioxidant³², immune-modulating³³ and signalling effects, the latter of which have been investigated in terms of Ppar activities in BR-treated mice. Results attest to an insulin-signalling effect of BR, ultimately modulating body weight, that appears to be in parts mediated through Ppar γ ²⁶.

Against this background, an observational case control study involving 120 healthy age- and gender-matched male and female subjects with and without GS, was conducted. The main aim was to further explain these striking metabolic differences, mainly reflecting in beneficial body composition, glucose- and lipid profile, as well as in apparently altered energetic regulations in response to fasting. To further explore particularities of metabolic regulation in GS, a molecular approach was used focusing on the AMPK pathway.

Figure 1. Comparison of measures of the AMPK pathway between the study groups (GS, C), including all subjects (male and female). Levels of (phosphorylated) proteins (AMPK $\alpha 1/\alpha 2$, Ppar α and γ , PgC 1 α) were analysed using the method of flow cytometry. Data are expressed as relative fluorescence units [rfU], and compared between subjects with Gilbert's syndrome (GS; *UGT1A1**28 promoter mutation), and controls (C). * Indicates significant differences between groups ($p \leq 0.05$). Medians can be found in Table 2. Abbreviations: pAMPK $\alpha 1/\alpha 2$: Phosphorylated 5'-AMP activated kinase; pPpar α : Phosphorylated peroxisome proliferator activated receptor alpha; pPpar γ : Phosphorylated peroxisome proliferator activated receptor gamma; PgC 1 α : Peroxisome proliferator-activated receptor c coactivator 1.

Results

Demographic and behavioural comparison between GS- and C subjects. Subjects between study groups did not significantly differ in terms of age distribution, or aspects of their lifestyles. As expected and crucial in terms of the study design, significant inter-group differences were found only for UCB and respective distribution of the number of TA-repeats (*i.e.* *UGT1A1**28 genotype). These genotype distributions reflect what has been previously reported^{34,35}. As for relevant differences in physical activity, male control (C) subjects (self-reportedly) were significantly more active than GS individuals (Table 1). There were no differences in parameters of liver health and iron status (AST, ALT, γ -GT, LDH, albumin, transferrin and ferritin) between GS and C individuals, as has been summarized by our group recently³⁶.

Comparison of metabolic parameters between GS- and C subjects. Biomarkers that were analysed in both study groups are summarized in Table 2 (including all subjects), Table 3 (for males only), and in Table 4 (for females only). Individual biomarkers were merged into groups, based on their respective effector-areas, namely energy-, carbohydrate- and lipid-metabolism.

Since exclusively healthy subjects were included in this study, all parameters reflecting metabolic health were within normal ranges throughout all groups (GS, C, males and females).

	Variable	Mean [°] (±sd)/median [^] (IQR)		<i>p</i> -value ^{*/T}
		GS	C	
<i>Biomarkers of energy metabolism</i>	pAMPK α1/α2 [rfU] [^]	187 (94) n = 58	112 (43) n = 56	0.000*
	pPpar α [rfU] [°]	400 (±88) n = 57	292 (±80) n = 58	0.000*
	pPpar γ [rfU] [°]	1401 (±398) n = 58	1086 (±378) n = 58	0.000*
	PgC 1α [rfU] [^]	206 (56) n = 58	165 (55) n = 58	0.000*
	AMPK α1 expr. [RQ] [^]	0.76 (0.25) n = 60	0.80 (0.25) n = 60	0.592
	Sirt-1 [ng/mL] [^]	3.16 (1.64) n = 47	2.92 (1.18) n = 52	0.491
	FGF-21 [μg/mL] [^]	0.35 (0.43) n = 52	0.10 (0.43) n = 51	0.086 ^T
	TSH [μU/mL] [^]	1.8 (1.2) n = 57	2.0 (1.2) n = 59	0.404
	T3 [pg/mL] [°]	3.24 (±0.41) n = 58	3.22 (±0.39) n = 59	0.844
	T4 [ng/dL] [^]	1.29 (0.20) n = 58	1.26 (0.20) n = 59	0.105
	BMI [kg/m ²] [°]	22.8 (±3.0) n = 60	25.4 (±4.9) n = 60	0.001*
	LBM [%] [^]	78.5 (8) n = 59	76.2 (13) n = 58	0.105
<i>Biomarkers of carbohydrate metabolism</i>	Glucose [mg/dL] [^]	81 (7) n = 59	86 (11) n = 60	0.004*
	HbA1c [%] [^]	5.0 (0.4) n = 60	5.1 (0.5) n = 60	0.116
	C-Peptide [ng/mL] [^]	1.2 (0.6) n = 58	1.4 (1) n = 59	0.004*
	Insulin [μU/mL] [^]	4.1 (2.5) n = 55	5.9 (4.7) n = 59	0.001*
<i>Biomarkers of lipid metabolism</i>	TChol [mg/dL] [^]	171 (47) n = 58	182 (67) n = 60	0.112
	HDL Cholesterol [mg/dL] [^]	62 (25) n = 58	63 (27) n = 60	0.726
	LDL Cholesterol [mg/dL] [^]	93 (35) n = 58	100 (64) n = 60	0.112
	LDL/HDL ratio [^]	1.4 (0.7) n = 58	1.6 (1.2) n = 60	0.162
	TG [mg/dL] [^]	73 (31) n = 57	85 (50) n = 60	0.045*
	LPA2 [mg/dL] [^]	17 (21) n = 54	13 (18) n = 55	0.082 ^T
	ApoA1 [mg/dL] [°]	149 (±27) n = 60	152 (±30) n = 60	0.638
	ApoB [mg/dL] [°]	83 (±24) n = 60	90 (±28) n = 60	0.194
	Apo B/A1 ratio [^]	0.5 (0.2) n = 59	0.6 (0.3) n = 60	0.409

Table 2. Energy-, carbohydrate and lipid metabolic biomarkers of all subjects of the BiliHealth study.

(Table 2 summarises and compares (GS versus C) biomarkers of energy-, carbohydrate and lipid metabolism of all subjects of the BiliHealth study. Values are specified as applies according to distribution of data. For parametric variables, means[°] ± sd are shown, for non-parametric data, medians[^] (50th percentiles) and inter-quartile range (IQR) are displayed. Comparison of means for parametric data or of ranks (for non-parametric data) was completed using independent samples t-test or Mann-Whitney-U-test. P-value*: significant on a 5% level of significance; p-value^T: trend on a 10% level of trend. Abbreviations: GS: Gilbert's syndrome; C: Controls; pAMPK α1/α2: Phosphorylated 5'-AMP activated kinase; pPpar α: Phosphorylated peroxisome proliferator activated receptor alpha; pPpar γ: Phosphorylated peroxisome proliferator activated receptor gamma; PgC 1α: Peroxisome proliferator-activated receptor c coactivator 1; AMPK α1 expr.: AMPK gene expression as relative quantification, RQ to cDNA pool; Sirt-1: Sirtuin-1; FGF-21: Fibroblast growth factor 21; TSH: Thyroid stimulating hormone; T3: Free triiodothyronine; T4: Free thyroxine; BMI: Body mass index; LBM: Lean body mass; HbA1c: Glycated haemoglobin A1c; TChol: Total cholesterol; HDL: High density lipoprotein cholesterol; LDL: Low density lipoprotein cholesterol; TG: Triglyceride; LPA2: Lipoprotein A2; ApoA1: Apolipoprotein A1; ApoB: Apolipoprotein B.

In the subsequent paragraphs, features of metabolic health are described, summarized for the entire study population, as well as split into gender groups.

Biomarkers associated with energy metabolism in GS- versus C subjects. *All subjects.* Median phosphorylation/protein expression of AMPK α1/α2 and of associated downstream transcription factors (pPpar α, pPpar γ, PgC 1α; measured in PBMCs) was significantly higher in GS subjects versus controls (Fig. 1; *p* = 0.000). No group difference, however, was found in terms of *AMPKα1* gene expression (Table 2).

A trend towards higher FGF-21 serum concentration was found in GS relative to controls. No statistical difference was stated concerning Sirt-1 levels between groups, although values were slightly higher in GS as compared to controls.

Data on body composition differed between the groups, in that BMI levels were significantly lower in GS subjects relative to controls (*p* = 0.001), and lean body mass (LBM) was higher in GS, however, did not reach statistical significance.

Male subjects. Those significant differences in AMPK α1/α2 phosphorylation (and the downstream parameters pPpar α, pPpar γ, PgC 1α) found for both genders, were retained in males (*p* = 0.000, *p* = 0.003). Again, *AMPK1α* gene expression did not differ significantly between GS and C groups (Table 3).

Anthropometric measures were significantly different only in terms of BMI (*p* = 0.023), which was lower in the GS group. For males, no significant results were obtained for LBM.

MALES	Variable	Mean [°] (±sd)/median [^] (IQR)		p-value ^{°/T}
		GS	C	
<i>Biomarkers of energy metabolism</i>	pAMPK $\alpha 1/\alpha 2$ [rfU] [^]	201 (91) n = 39	107 (39) n = 37	0.000*
	pPpar α [rfU] [°]	385 (±84) n = 38	296 (±90) n = 39	0.000*
	pPpar γ [rfU] [^]	1384 (365) n = 39	1049 (438) n = 39	0.003*
	PgC 1 α [rfU] [^]	205 (53) n = 39	158 (56) n = 39	0.000*
	AMPK $\alpha 1$ expr. [RQ] [^]	0.76 (0.24) n = 40	0.78 (0.25) n = 40	0.729
	Sirt-1 [ng/mL] [^]	3.26 (1.66) n = 31	3.24 (1.77) n = 34	0.887
	FGF-21 [μ g/mL] [^]	0.33 (0.38) n = 34	0.13 (0.41) n = 34	0.147
	TSH [μ U/mL] [°]	1.8 (±0.8) n = 38	1.9 (±0.9) n = 39	0.372
	T3 [pg/mL] [°]	3.37 (±0.32) n = 39	3.31 (±0.39) n = 39	0.476
	T4 [ng/dL] [^]	1.30 (0.20) n = 39	1.27 (0.20) n = 39	0.163
	BMI [kg/m ²] [^]	22.5 (3.6) n = 40	24.9 (5.9) n = 40	0.023*
	LBM [%] [^]	79.4 (7.0) n = 40	80.9 (9.0) n = 38	0.838
<i>Biomarkers of carbohydrate metabolism</i>	Glucose [mg/dL] [°]	81 (7) n = 39	86 (11) n = 40	0.016*
	HbA1c [%] [^]	5.0 (0.4) n = 40	5.1 (0.5) n = 40	0.150
	C-Peptide [ng/mL] [^]	1.2 (0.7) n = 39	1.3 (1.1) n = 39	0.009*
	Insulin [μ U/mL] [^]	3.9 (2.4) n = 36	6.1 (5.8) n = 39	0.001*
<i>Biomarkers of lipid metabolism</i>	TChol [mg/dL] [^]	165 (49) n = 39	180 (71) n = 40	0.344
	HDL Chol [mg/dL] [°]	58 (±15) n = 39	57 (±14) n = 40	0.761
	LDL Chol [mg/dL] [^]	94 (36) n = 39	109 (70) n = 40	0.410
	LDL/HDL ratio [^]	1.6 (1.0) n = 39	1.8 (1.3) n = 40	0.315
	TG [mg/dL] [^]	73 (31) n = 38	95 (56) n = 40	0.023*
	LPA2 [mg/dL] [^]	22 (27) n = 37	14 (21) n = 37	0.048*
	ApoA1 [mg/dL] [°]	140 (±24) n = 40	141 (±24) n = 40	0.882
	ApoB [mg/dL] [^]	84 (33) n = 40	82 (57) n = 40	0.939
	Apo B/A1 ratio [^]	0.6 (0.2) n = 39	0.6 (0.3) n = 40	0.521

Table 3. Energy-, carbohydrate and lipid metabolic biomarkers of male subjects of the BiliHealth study.

(Table 3 summarises and compares (GS versus C) biomarkers of energy-, carbohydrate and lipid metabolism of male subjects of the BiliHealth study. Values are specified as applies according to distribution of data. For parametric variables, means[°] (±sd) are shown, for non-parametric data, medians[^] (50th percentiles) and inter-quartile range (IQR) are displayed. Comparison of means (for parametric data) or of ranks (for non-parametric data) was completed using independent samples t-test or Mann-Whitney-U-test (*p-value: significant on a 5% level of significance; ^Tp-value: trend on a 10% level of trend). Abbreviations: GS: Gilbert's syndrome; C: Controls; pAMPK $\alpha 1/\alpha 2$: Phosphorylated 5'-AMP activated kinase; pPpar α : Phosphorylated peroxisome proliferator activated receptor alpha; pPpar γ : Phosphorylated peroxisome proliferator activated receptor gamma; PgC 1 α : Peroxisome proliferator-activated receptor c coactivator 1; AMPK $\alpha 1$ expr.: AMPK gene expression as relative quantification, RQ to cDNA pool; Sirt-1: Sirtuin-1; FGF-21: Fibroblast growth factor 21; TSH: Thyroid stimulating hormone; T3: Free triiodothyronine; T4: Free thyroxine; BMI: Body mass index; LBM: Lean body mass; HbA1c: Glycated haemoglobin A1c; TChol: Total cholesterol; HDL: High density lipoprotein cholesterol; LDL: Low density lipoprotein cholesterol; TG: Triglyceride; LPA2: Lipoprotein A2; ApoA1: Apolipoprotein A1; ApoB: Apolipoprotein B.

Female subjects. Phosphorylation of AMPK $\alpha 1/\alpha 2$ and its downstream effectors (pPpar α , pPpar γ , PgC 1 α) was significantly higher in female GS versus C subjects ($p = 0.037$, $p = 0.000$). In summary, this result is retained throughout the gender groups. For *AMPK $\alpha 1$* gene expression, again no significant results were found (Table 4).

In terms of body composition, both BMI and LBM differed significantly between the groups ($p = 0.017$, $p = 0.011$), in that BMI was lower and LBM was higher in female GS versus C.

Biomarkers associated with carbohydrate metabolism in GS- versus C subjects. *All subjects.* When considering the entire study population, fasting plasma glucose levels as well as concentrations of insulin and C-peptide, were significantly lower in the GS group, as compared to controls ($p = 0.004$, $p = 0.001$) (Table 2).

Male subjects. Again, as stated for the entire study group, fasting glucose levels were significantly lower in male GS versus C, as were insulin and C-peptide concentrations ($p = 0.016$, $p = 0.009$, $p = 0.001$) (Table 3).

Female subjects. Differences in parameters of glucose metabolism did not reach statistical significance between the study groups (Table 4).

Biomarkers associated with lipid metabolism in GS- versus C subjects. *All subjects.* Plasma TG levels were significantly lower in GS subjects ($p = 0.045$), LPA2 by trend was higher in GS. The remaining lipid parameters (as listed in Table 2) did not differ significantly between the groups (Table 2).

FEMALES	Variable	Mean [°] (±sd)/median [^] (IQR)		p-value ^{*/T}
		GS	C	
<i>Biomarkers of energy metabolism</i>	pAMPK $\alpha 1/\alpha 2$ [rfU] [°]	173 (±59) n = 19	135 (±49) n = 19	0.037*
	pPpar α [rfU] [°]	431 (±90) n = 19	283 (±58) n = 19	0.000*
	pPpar γ [rfU] [°]	1543 (±351) n = 19	1069 (±376) n = 19	0.000*
	PgC 1 α [rfU] [°]	221 (±41) n = 19	167 (±33) n = 19	0.000*
	AMPK $\alpha 1$ expr. [RQ] [^]	0.81 (0.4) n = 20	0.85 (0.35) n = 20	0.303
	Sirt-1 [ng/mL] [^]	2.79 (1.74) n = 16	2.67 (0.96) n = 18	0.731
	FGF-21 [μ g/mL] [^]	0.46 (0.65) n = 18	0.25 (0.67) n = 17	0.428
	TSH [μ U/mL] [^]	1.9 (1.6) n = 19	2.1 (1.2) n = 20	0.747
	T3 [pg/mL] [°]	2.96 (±0.45) n = 19	3.04 (±0.35) n = 20	0.530
	T4 [ng/dL] [°]	1.25 (±0.15) n = 19	1.20 (±0.17) n = 20	0.371
	BMI [kg/m ²] [°]	21.8 (±2.7) n = 20	24.7 (±4.4) n = 20	0.017*
	LBM [%] [°]	75.6 (±7.0) n = 19	69.2 (±7.8) n = 20	0.011*
<i>Biomarkers of carbohydrate metabolism</i>	Glucose [mg/dL] [^]	81 (8) n = 20	85 (10) n = 20	0.136
	HbA1c [%] [°]	5.0 (±0.5) n = 20	5.1 (±0.4) n = 20	0.379
	C-Peptide [ng/mL] [^]	1.3 (0.5) n = 19	1.5 (0.8) n = 20	0.146
	Insulin [μ U/mL] [^]	4.3 (2.6) n = 19	4.7 (4.9) n = 20	0.261
<i>Biomarkers of lipid metabolism</i>	TChol [mg/dL] [°]	179 (±33) n = 19	194 (±33) n = 20	0.154
	HDL Cholesterol [mg/dL] [°]	81 (±19) n = 19	76 (±20) n = 20	0.476
	LDL Cholesterol [mg/dL] [°]	84 (±32) n = 19	103 (±31) n = 20	0.071 ^T
	LDL/HDL ratio [°]	1.1 (±0.6) n = 19	1.5 (±0.7) n = 20	0.099 ^T
	TG [mg/dL] [^]	70 (34) n = 19	66 (33) n = 20	0.725
	LPA2 [mg/dL] [^]	14.0 (12.0) n = 17	10.5 (17.0) n = 18	0.865
	ApoA1 [mg/dL] [°]	168 (±23) n = 20	174 (±28) n = 20	0.485
	ApoB [mg/dL] [°]	77 (±21) n = 20	89 (±23) n = 20	0.087 ^T
	Apo B/A1 ratio [^]	0.5 (0.2) n = 20	0.5 (0.2) n = 20	0.463

Table 4. Energy-, carbohydrate and lipid metabolic biomarkers of female subjects of the BiliHealth study.

(Table 4 summarises and compares (GS versus C) biomarkers of energy-, carbohydrate and lipid metabolism of female subjects of the BiliHealth study. Values are specified as applies according to distribution of data. For parametric variables, means[°] (±sd) are shown, for non-parametric data, medians[^] (50th percentiles) and inter-quartile range (IQR) are displayed. Comparison of means (for parametric data) or of ranks (for non-parametric data) was completed using independent samples t-test or Mann-Whitney-U-test (*p-value: significant on a 5% level of significance; ^Tp-value: trend on a 10% level of trend). Abbreviations: GS: Gilbert's syndrome; C: Controls; pAMPK $\alpha 1/\alpha 2$: Phosphorylated 5'-AMP activated kinase; pPpar α : Phosphorylated peroxisome proliferator activated receptor alpha; pPpar γ : Phosphorylated peroxisome proliferator activated receptor gamma; PgC 1 α : Peroxisome proliferator-activated receptor c coactivator 1; AMPK $\alpha 1$ expr.: AMPK gene expression as relative quantification, RQ to cDNA pool; Sirt-1: Sirtuin-1; FGF-21: Fibroblast growth factor 21; TSH: Thyroid stimulating hormone; T3: Free triiodothyronine; T4: Free thyroxine; BMI: Body mass index; LBM: Lean body mass; HbA1c: Glycated haemoglobin A1c; TChol: Total cholesterol; HDL: High density lipoprotein cholesterol; LDL: Low density lipoprotein cholesterol; TG: Triglyceride; LPA2: Lipoprotein A2; ApoA1: Apolipoprotein A1; ApoB: Apolipoprotein B.

Male subjects. Those significant results reported above for both genders in terms of plasma TG and LPA2, were in retained in male subjects. LPA2 was again higher and TG levels were significantly lower in male GS ($p = 0.048$, $p = 0.023$). For the remaining parameters (as listed in Table 3), no significant results were obtained (Table 3).

Female subjects. As for the plasma lipid fractions, results for females are different to those presented for males. In female GS, LDL was lower by trend relative to C, as was the LDL/HDL ratio. Also Apo B trended to be lower in female GS as compared to controls (Table 4).

Inter-variable connection of metabolic parameters and the AMPK-pathway, is merely established via UCB and the underlying UGT1A1 genotype (-TA repeats). For the reason of statistical validity and power, the entire dataset was analysed for inter-variable connections. A graphical summary of the correlations found is presented in Fig. 2, generating an idea as to how the parameters analysed could be networking on a physiological level. A detailed list of significant inter-variable associations, expressed as correlation coefficients (R) with corresponding p-values (in brackets), is provided in the figure.

To begin with UCB as the crucial determinant of the study design, and main feature of GS, strong correlations were found with pAMPK $\alpha 1/\alpha 2$, ($R = 0.507$, $p = 0.000$), and with downstream transcription factors (Fig. 2). A similar association was found when looking at the *UGT1A1* genotype, representing the underlying molecular background of GS ($R = 0.412$, $p = 0.000$). Gender-specific associations between UCB/*UGT1A1* and AMPK regulations, are visualized in Fig. 3. Interestingly, stronger correlations were generally found in women than in men.

Figure 2. Graphical summary of inter-variable connections. Figure 2 illustrates statistical connections of variables of interest. Bivariate correlations were calculated for the entire study population using the model of Spearman's rho. R coefficients and p-values ($p \leq 0.05$; in brackets) are presented in the grey box and summarised in a graphical model. For ease of reading, lifestyle factors have not been incorporated in detail, and are therefore only abstracted (bottom of figure). Abbreviations: UCB: unconjugated bilirubin; *UGT1A1*: *UGT1A1* genotype; pAMPK α1/α2: Phosphorylated 5'-AMP activated kinase; pPpar α: Phosphorylated peroxisome proliferator activated receptor alpha; pPpar γ: Phosphorylated peroxisome proliferator activated receptor gamma; PgC 1α: Peroxisome proliferator-activated receptor c coactivator 1; Sirt-1: Sirtuin-1; FGF-21: Fibroblast growth factor 21; T3: Free triiodothyronine; BMI: Body mass index; LBM: Lean body mass; HbA1c: Glycated haemoglobin A1c; TChol: Total cholesterol; HDL: High density lipoprotein cholesterol; LDL: Low density lipoprotein cholesterol; TG: Triglyceride; LPA2: Lipoprotein A2; ApoA1: Apolipoprotein A1; ApoB: Apolipoprotein B.

Not surprisingly, UCB levels and *UGT1A1* genotype have been found to be strongly correlated ($R = 0.731$, $p = 0.000$). For these two specific features of GS (UCB and *UGT1A1* genotype), negative correlations were observed for a series of important lipid- and glucose biomarkers (as are listed in Fig. 2), emphasizing an improved metabolic state as UCB levels/TA-repeats increase.

The same negative connection applies to UCB/*UGT1A1* genotype and the anthropometric measure of BMI ($R = -0.274$, $p = 0.002$), whereas conversely, a positive association was found with LBM ($R = 0.217$, $p = 0.019$). Pursuing the interplay of anthropometric measures and other parameters, revealed connections of BMI and LBM with markers of lipid and carbohydrate metabolism (Fig. 2). Furthermore, an association of LBM with Sirt-1 was found ($R = 0.242$, $p = 0.017$), the latter of which statistically looping back to carbohydrate metabolism (HbA1c; $R = -0.235$, $p = 0.019$). HbA1c was also negatively associated with Ppar γ ($R = -0.204$, $p = 0.028$), one of the downstream effectors of the AMPK pathway, emphasizing its direct connection to energy- and carbohydrate metabolism (Fig. 2).

In summary, those correlations found point to close connections between characterising features of GS, body composition and an altered metabolic state in this condition, altogether having strong implications for macronutrient metabolism (carbohydrate and lipid), and importantly for energy turnover.

Heat maps visualizing correlated variables with respect to energy-, glucose- and lipid metabolism can be found on the online supplementary information (supplementary Figures S2 a, b, c).

Figure 3. Correlations of UCB (a–d) and the UGT1A1 genotype (e–h), with measures of the AMPK pathway. Figure 3 illustrates gender-specific correlations of UCB and the UGT1A1 genotype with measures of the AMPK pathway. Bivariate correlations between UCB/UGT1A1 genotype (-TA repeats: 6/6 controls, 6/7 heterozygous, 7/7 Gilbert's syndrome) and measures of the AMPK pathway were calculated for each gender (m = male, f = female), using the model of Spearman's rho. R coefficients and p-values ($p \leq 0.05$; in brackets) are as follows: **UCB * pAMPK $\alpha 1/\alpha 2$** : m 0.594 (0.000); f 0.255 (0.122), **UCB * PgC1 α** : m 0.376 (0.001); f 0.467 (0.003), **UCB * pPpar α** : m 0.435 (0.000); f 0.575 (0.000), **UCB * pPpar γ** : m 0.354 (0.001); f 0.324 (0.047), **UGT1A1 * pAMPK $\alpha 1/\alpha 2$** : m 0.541 (p 0.000); f 0.156 (p 0.362), **UGT1A1 * PgC1 α** : m 0.265 (p 0.023); f 0.551 (p 0.001), **UGT1A1 * Ppar α** : m 0.365 (p 0.002); f 0.661 (p 0.000), **UGT1A1 * Ppar γ** : m 0.191 (p 0.023); f 0.435 (p 0.008). Abbreviations: UCB: unconjugated bilirubin; pAMPK $\alpha 1/\alpha 2$: Phosphorylated 5'-AMP activated kinase; pPpar α : Phosphorylated peroxisome proliferator activated receptor alpha; pPpar γ : Phosphorylated peroxisome proliferator activated receptor gamma; PgC1 α : Peroxisome proliferator-activated receptor c coactivator 1; WT: wild type (control subjects); GS: Gilbert's syndrome.

It all starts with UCB: bilirubin has statistical explanatory power for AMPK pathway regulations and body composition. To further pursue inter-variable connections, and to explore possibilities as to how those entities studied could explain each other, stepwise linear regression models were generated. A graphical abstract of the most important findings, can be found in Fig. 4, where percentages (based on corrected R^2 regression coefficients) specifying inter-variable explanatory power, are presented. Furthermore, tables summarising all relevant correlations that were found, are provided (Tables 5 and 6).

Most compelling, and in line with the findings from bivariate correlation analysis, UCB has noteworthy explanatory power for AMPK phosphorylation (pAMPK, 19%), and, although not as pronounced, for related pathway characteristics (pPpar α , 2.8%). Importantly, UCB furthermore connects the AMPK-pathway with body composition, by in parts explaining the variable BMI (7.2%), and possibly providing an important explanation for the significantly lower BMI stated in GS subjects, relative to controls (Table 1).

Interestingly, however not surprisingly, measures of body composition (BMI, LBM) appear to be merely linked to parameters of lipid metabolism, thereby likely explaining the improved lipid status determined in GS versus control-subjects in this (Tables 2–4), and previous studies^{6,7}.

An entirely new inter-variable dependence was found, in that LBM had some explanatory power for Sirt-1 (6.9%), an important controller of metabolism with respect to ageing.

The measure of LBM was furthermore interlinked with PgC1 α , being the immediate activator of Ppar α and γ . This is clearly emphasized by its substantial explanatory power for the latter two (74.3% and 49.6%, respectively). Interestingly, there seems to be also an inverse correlation between these variables, suggesting a feedback-loop from Ppar α/γ to PgC1 α (78.1%).

As is known from the literature^{37,38}, and newly reported here for the matrix of PBMCs, the AMPK pathway *via* its effectors Ppar α and γ , is ultimately linked to glucose metabolism, thereby likely explaining the relatively improved glucose metabolism generally determined in GS subjects (Tables 2–4). On a larger scale, this result provides a mechanism for the known low prevalence of type II diabetes among subjects having GS⁸.

Measures of lifestyle (physical activity, frequency of consuming specific foods) did not have significant further influence on variables of the AMPK pathway, as has been confirmed using regression analysis.

Discussion

The aim of this study was to establish a theory to explain the compelling differences repeatedly found between GS and control subjects, concerning body composition and overall metabolic health. The AMPK pathway was

DEPENDENT VARIABLES:	pAMPK $\alpha 1/\alpha 2$	PgC1 α	Ppar α	Ppar γ	AMPK $\alpha 1$ expr.	UCB	UGT1A1 genotype	BMI	LBM	T3
pAMPK $\alpha 1/\alpha 2$						0.190 (0.000)				
PgC1 α	0.070 ^{GS} (0.006)		0.743 (0.000)	0.781 (+Ppar α) (0.000)	0.060 (0.041)				0.083 ^{GS} (0.020)	
Ppar α		0.743 (0.000)					0.771 (+PgC1 α) (0.000)			
Ppar γ		0.496 (0.000)								
AMPK $\alpha 1$ expr.										
BMI						0.072 (0.004)				
LBM								0.483 (0.000)		0.538 (+BMI) (0.000)
Sirt-1									0.069 (0.009)	

Table 5. Stepwise linear regression analysis for key variables included in Fig. 2. Corrected R^2 coefficients and corresponding p-values (in brackets) from stepwise linear regression analysis are provided. Unspecified regressions are valid for the entire study population. ^ceffect valid for control subjects only; ^{GS}effect valid for GS subjects only. Abbreviations: pAMPK $\alpha 1/\alpha 2$: Phosphorylated 5'-AMP activated kinase; PgC1 α : Peroxisome proliferator-activated receptor c coactivator 1; pPpar α : Phosphorylated peroxisome proliferator activated receptor alpha; Ppar γ : Phosphorylated peroxisome proliferator activated receptor gamma; AMPK $\alpha 1$ expr.: AMPK $\alpha 1$ gene expression; UCB: unconjugated bilirubin; UGT1A1: UDP glucuronosyltransferase 1A1 polymorphism; BMI: Body mass index; LBM: Lean body mass; Sirt-1: Sirtuin-1; T3: Free triiodothyronine. Included variables: **pAMPK $\alpha 1/\alpha 2$** : AMPK1a expr., PgC1 α , pPpar α , pPpar γ , TChol, TG, UCB, UGT1A1 genotype, BMI, LBM, LDL/HDL, TSH, T3, T4. **PgC1 α** : pAMPK $\alpha 1/\alpha 2$, AMPK1a expr., Ppar α , Ppar γ , TChol, TG, UCB, UGT1A1 genotype, BMI, LBM, LDL/HDL, TSH, T3, T4. **pPpar α** : pAMPK $\alpha 1/\alpha 2$, AMPK1a expr., Ppar γ , TChol, TG, UCB, UGT1A1 genotype, BMI, LBM, LDL/HDL, TSH, T3, T4. **AMPK $\alpha 1$ expr.**: pAMPK $\alpha 1/\alpha 2$, Ppar α , Ppar γ , TChol, TG, UCB, UGT1A1 genotype, BMI, LBM, LDL/HDL, TSH, T3, T4. **BMI**: AMPK $\alpha 1$ expr., pAMPK $\alpha 1/\alpha 2$, PgC1 α , Ppar α , Ppar γ , UCB, UGT1A1 genotype, LBM, TSH, T3, T4. **LBM**: AMPK $\alpha 1$ expr., pAMPK $\alpha 1/\alpha 2$, PgC1 α , Ppar α , Ppar γ , UCB, UGT1A1 genotype, BMI, TSH, T3, T4. **Sirt-1**: AMPK $\alpha 1$ expr., pAMPK $\alpha 1/\alpha 2$, PgC1 α , Ppar α , Ppar γ , UCB, UGT1A1 genotype, BMI, LBM, TSH, T3, T4, FGF21, TChol, TG, LDL/HDL. (In further analyses the variables age, gender and those specifying lifestyle were included, however these procedures did not significantly change the models' outcome).

the investigated model of choice, elegantly networking and determining features of energy- and macronutrient metabolism on a cellular and whole-body level. Specific findings of the study at hand are discussed and summarized in the following paragraphs. The obtained results provide robust evidence, that the energy- and macronutrient metabolic response to fasting are clearly boosted in GS. Accordingly, even though all subjects were metabolically healthy and within the reference ranges concerning their blood biochemistry parameters, several inter-group differences were found, confirming the improved metabolic health status of GS individuals. The relative extent to which these metabolic shifts are due to a direct effect of elevated UCB levels or based on a more complex genetic association with the *UGT1A1**28 promoter mutation, remains to be clarified.

As stated in the materials and methods section, all 120 subjects were required to fast on the day before blood sampling (400 kcal restriction), as well as overnight for 16 (± 1) hours. This is important to mention since serum UCB is known to rise in response to fasting, and was found to be the main determining factor of post-translational AMPK $\alpha 1/\alpha 2$ activation, as well as of BMI. This important feature of body composition was significantly lower in GS individuals.

AMPK – it all starts with physical stress. The tight association of UCB and AMPK $\alpha 1/\alpha 2$ activity is not surprising and was expected, since both effectors are triggered by fasting leading to a drop in ATP, and by other physical stressors including exercise. On a molecular level, these common influencing variables emphasize a potential molecular cross-talk, thereby connecting GS genetics with AMPK $\alpha 1/\alpha 2$ activation. The importance of an altered fasting response specifically in GS, is further highlighted by the fact that levels of *AMPK $\alpha 1$* gene expression were equal between the groups. With reference to the influencing factor of physical activity, control subjects (self-reportedly) more frequently engaged in physical activity. In theory this study group (C) should have thus benefitted from the various health-benefits known to arise from exercising, and AMPK activity should have been higher. However, the obtained results reported an opposing trend, in that AMPK phosphorylation was significantly higher in GS versus controls. Therefore AMPK $\alpha 1/\alpha 2$ was more re-/active, with significantly improved parameters of metabolic health, including glucose, C-peptide, insulin and TG in GS subjects. Interestingly the parameter of TG in fact was found to be the “universal” statistical connector and determinant of the pathways explored, in that it had explanatory powers for C-peptide, insulin, and all fractions of cholesterol (TChol, HDL, LDL).

AMPK from a different angle – metabolic interplay in GS. With reference to the observed phenomenon of lower plasma lipid fractions, it is well established that Ppar α and γ trigger pathways that are involved in lipogenesis and lipid storage^{39,40}. When activated through ligand-binding or phosphorylation (e. g. through

Figure 4. Abstract of inter-variable connection and dependence, based on regression analysis. Stepwise linear regression analysis was performed to assess inter-variable dependence and explanatory power (%), based on corrected R^2 values and a p-value of ≤ 0.05 (Tables 5 and 6). ^Ceffect found in control subjects only, ^{GS}effect found in GS subjects only, (All remaining unspecified correlations are valid for both study groups.), Abbreviations: UCB: unconjugated bilirubin; *UGT1A1*: *UGT1A1* genotype; *AMPKα1*: 5'-AMP activated kinase $\alpha 1$ gene expression; pAMPK $\alpha 1/\alpha 2$: Phosphorylated 5'-AMP activated kinase; pPpar α : Phosphorylated peroxisome proliferator activated receptor alpha; pPpar γ : Phosphorylated peroxisome proliferator activated receptor gamma; PgC 1 α : Peroxisome proliferator-activated receptor c coactivator 1; Sirt-1: Sirtuin-1; BMI: Body mass index; LBM: Lean body mass.

AMPK), these pathways lead to body-wide redistribution of fat, consequently lowering TG levels, and thereby improving insulin sensitivity. This effect was clearly present in male GS individuals, expressed through significantly lower TG levels, along with slightly lower TChol, and an improved glucose metabolism. This confirms the above role for Ppars as ultimate regulators of lipid metabolism (ultimately influencing that of glucose), and implies an even more pronounced metabolic effect in GS individuals, likely based on their more re-/active AMPK pathway. These important results readily connect to the low prevalence of metabolic diseases previously stated for GS individuals^{6,7}.

The significantly lower TChol levels previously stated for GS individuals⁵, and the slightly lower TChol levels reported for GS individuals in this study, could be predicated once more on this group's increased AMPK activity. AMPK has a known post-translational deactivating effect on HMG-CoA reductase (3-hydroxy-3-methyl-glutaryl-CoA reductase; not measured in this study), which is the rate-limiting enzyme in cholesterol synthesis⁴¹. This may be hypothesised to contribute to the slightly lower TChol levels in the GS group.

With reference to aspects of glucose regulation, a statistical association of Ppar γ activity (together with glucose), and the long term glucose parameter HbA1c was found. This is in agreement with results from animal studies in which an indirect BR-mediated insulin effect through the direct Ppar γ agonist HO-1²⁶, and an insulin-sensitizing activity of BR based on anti-inflammation⁴² have been reported. The overall effects and proposed underlying mechanisms were similarly found in the present study, in that GS individuals (like BR-treated mice in the above studies), were generally lighter, leaner and as mentioned, had improved glucose parameters relative to controls.

Key players in body composition – UCB, Ppars and anti-inflammation. Excess body mass is a known crucial regulator of glucose homeostasis, lipid metabolism^{43,44} and inflammation⁴⁵. An increased fat mass triggers inflammation, involving the production and release of TNF α and IL-6⁴⁵. Inversely, the comparably higher LBM that was determined in female GS individuals, as well as the lower BMI reported for all GS individuals, both would have an easing impact on inflammation. In the present study, it is evident that entities of metabolic health are apparently influenced to a large extent by BMI. This marker was in part explained by UCB levels, thereby connecting AMPK activity with body composition (Fig. 4). C-reactive protein (CRP), as well as TNF α and IL-6 were in fact (significantly) lower in GS individuals as compared to controls (results presented in ref. 36). This emphasizes the above hypothesis of a possible involvement of BR (UCB in the present study), in lowering inflammation through an increase in energy turnover (AMPK phosphorylation), and provides a link to improved body composition. Considering the fact that all study participants were free of (inflammatory) diseases, this result is particularly remarkable.

DEPENDENT VARIABLES:	C-Peptide	Glucose	TChol	TG	UCB	pPpar γ	LBM	BMI
HbA1c		0.314 (0.000)				0.340 (+Glucose) (0.000)		
C-Peptide		0.575 (+TG, BMI) (0.000)		0.447 (0.000)				0.544 (+TG) (0.000)
Insulin				0.350 (0.000)				0.415 (+TG) (0.000)
Glucose	0.264 (0.000)							
TChol				0.158 (0.000)			0.205 (+TG) (0.000)	
HDL				0.187 (0.000)				
LDL				0.198 (0.000)				0.226 (+TG) (0.000)
TG			0.318 (+BMI) (0.000)					0.249 (0.000)
Apo A1							0.190 (+BMI) (0.000)	0.047 (0.018)
Apo B				0.274 (0.000)				0.303 (+TG) (0.000)
LPA2					0.034 (0.043)			

Table 6. Stepwise linear regression analysis with reference to metabolic pathways (Fig. 2). Corrected R² coefficients and p-values (in brackets) from stepwise linear regression analysis, involving the entire study population, are provided. HbA1c: Glycated haemoglobin A1c; pPpar γ : Phosphorylated peroxisome proliferator activated receptor gamma; T3: Free triiodothyronine; T4: Free thyroxine; BMI: Body mass index; LBM: Lean body mass; TChol: Total cholesterol; HDL: High density lipoprotein cholesterol; LDL: Low density lipoprotein cholesterol; TG: Triglyceride; LPA2: Lipoprotein A2; ApoA1: Apolipoprotein A1; ApoB: Apolipoprotein B. Included variables: **HbA1c**: UCB, TSH, T3, T4, UGT1A1-genotype, pAMPK α 1/ α 2, PgC1 α , pPpar α , pPpar γ , AMPK α 1 expr., BMI, LBM, TG, FGF-21, C-Peptide, Glucose. **C-Peptide/Insulin**: UCB, TSH, T3, T4, UGT1A1-genotype, pAMPK α 1/ α 2, pPpar α , pPpar γ , PgC1 α , AMPK α 1 expr., BMI, LBM, TG, FGF-21, Glucose, HbA1c. **Glucose**: UCB, TSH, T3, T4, UGT1A1-genotype, pAMPK α 1/ α 2, pPpar α , pPpar γ , PgC1 α , AMPK α 1 expr., BMI, LBM, TG, FGF-21, C-Peptide, TChol, LDL/HDL. **TChol**: UCB, TSH, T3, T4, UGT1A1-genotype, pAMPK α 1/ α 2, pPpar α , pPpar γ , PgC1 α , AMPK α 1 expr., BMI, LBM, TG, FGF-21. **HDL**: UCB, TSH, T3, T4, UGT1A1-genotype, pAMPK α 1/ α 2, pPpar α , pPpar γ , PgC1 α , AMPK α 1 expr., BMI, LBM, TG, FGF-21. **LDL**: UCB, TSH, T3, T4, UGT1A1-genotype, pAMPK α 1/ α 2, pPpar α , pPpar γ , PgC1 α , AMPK α 1 expr., BMI, LBM, TG, FGF-21. **TG**: UCB, TSH, T3, T4, UGT1A1-genotype, pAMPK α 1/ α 2, pPpar α , pPpar γ , PgC1 α , AMPK α 1 expr., BMI, LBM, TG, FGF-21, TChol, Glucose, HbA1c. **Apo A1**: UCB, TSH, T3, T4, UGT1A1-genotype, pAMPK α 1/ α 2, pPpar α , pPpar γ , PgC1 α , AMPK α 1 expr., BMI, LBM, TG, FGF-21. **Apo B**: UCB, TSH, T3, T4, UGT1A1-genotype, pAMPK α 1/ α 2, pPpar α , pPpar γ , PgC1 α , AMPK α 1 expr., BMI, LBM, TG, FGF-21. **LPA2**: UCB, TSH, T3, T4, UGT1A1-genotype, pAMPK α 1/ α 2, pPpar α , pPpar γ , PgC1 α , AMPK α 1 expr., BMI, LBM, TG, FGF-21. (In further analyses the variables age, gender and those specifying lifestyle were included, however these procedures did not significantly change the models' outcome).

With reference to body composition another interesting observation was made in that LBM had explanatory power for Sirt-1, a known controller of metabolism with specific relevance to ageing. This result could further bridge the gap towards explaining the epidemiological evidence for longevity in GS, which has been experimentally explored recently³⁶.

More detailed statistical analyses into body composition and its connection to energy metabolism revealed an interesting gender-specific effect, which to date cannot be definitively explained. It is, however, possibly based on the gender-specific difference in oestrogen levels, that are known to influence energy metabolic pathways^{46,47}. As mentioned, LBM was significantly higher in GS individuals (relative to controls) only in females, and the beneficial difference in BMI between the two female groups was more pronounced as compared to that between the male groups (GS versus C). Ultimately connecting these results to energy turnover, they are readily confirmed by the generally stronger correlations between the AMPK pathway and increasing -TA repeats and UCB levels, found in women as compared to men (Fig. 3a,b). These observations are particularly remarkable in view of the relatively smaller female versus male group sizes.

Summary

In conclusion, the AMPK pathway not only is a master regulator of (energy) metabolism and main crossroad of various pathways, it furthermore seems to be a powerful switch that in GS more readily reacts to fasting, possibly leading to an increased energy turnover in this condition.

In this study, (i) not only those beneficial metabolic features were confirmed that had been established previously for GS individuals, but (ii) also the new finding of an apparently boosted AMPK pathway in GS in response to fasting, was presented for the first time. To this end, it cannot be estimated to which extent a potential increase in energy turnover in GS individuals is based on adaptive thermo- and mitogenesis, and possibly adipose tissue browning.

However, our findings expand and complement data previously obtained from animal studies^{26,42}, and propose a precise connection point to pursue future investigations into the molecular background and particularities of metabolic regulation in GS. Specific approaches could include immune-precipitation analyses to assess potential direct binding of BR as Ppar agonist, or microarray/genome wide association studies (GWAS) to screen for SNP-SNP associations or for SNP-interactions with certain phenotypic characteristics and aspects of metabolism in GS.

Materials and Methods

Subjects and study design. This study (abbreviated “BiliHealth”) was designed as an observational case-control study, at a single centre in Vienna, Austria. The study was performed at the Department of Clinical Pharmacology at Vienna General Hospital, and subjects were recruited between June 2014 and January 2015, by direct advertising (bulletin boards, posters and flyers) and from the department's subject database.

One hundred twenty-eight (128) healthy subjects between 20 and 80 years of age were initially recruited from the general Austrian population. Eight thereof, had to be excluded for medical reasons. Exclusion criteria included smoking, excess drinking, routine intake of medications and nutritional supplements, pregnancy, acute and chronic (inflammatory/metabolic) diseases, liver diseases, present or past neoplasia and organ transplants. After providing their signed written consent form, each subject completed an initial health check-up (fasting blood biochemistry including levels of unconjugated bilirubin (UCB) and liver enzymes, blood pressure, body weight/-height, questionnaires).

A total of 80 males and 40 females completed the study. This gender distribution is representative of the occurrence of GS in the general population⁴⁸. All subjects were age- and gender-matched, and study group allocation (GS, C) was based on the subjects' respective fasting serum UCB concentrations ($</\geq 17.1 \mu\text{M}$)⁴⁸, that had been analysed using HPLC. For the most part, subjects with GS (in contrast to C) showed visible signs of mild jaundice, reflecting in a yellowish pigmentation of the skin and the conjunctival membranes over the sclerae. Liver parameters and parameters of haemolysis were within the normal ranges. Participants were furthermore allocated to age groups ($</\geq 35$ years of age). For a graphical summary of the study design refer to supplementary Figure S1.

For the purpose of diagnosing GS, all subjects of both study groups were required to fast on the day before participating in the study, and therefore had to follow a 400 kcal fasting protocol^{4,49}. Furthermore, a complete overnight fast of 16 (± 1) hours was required, before the day of blood sampling.

Characteristics of the study population, including age distribution, UCB levels and aspects of lifestyle, are summarized in Table 1.

Ethics. This study was approved by the Ethics Commission of the Medical University of Vienna (No. 1164/2014), and was conducted in accordance with the approved guidelines by the Declaration of Helsinki.

Blood biochemistry (whole blood, plasma, serum). For each subject, fasting blood samples were collected on a single occasion (baseline), no longer than two weeks from the entry health check-up. Samples were drawn by venepuncture into EDTA, Li-Heparin and serum tubes (K₂EDTA, Li-Heparin and Z Serum Sep, respectively). Samples were cooled and protected from light until being analysed or aliquoted. Aliquots were stored at -80°C until further analysis.

Besides UCB, liver enzymes (aspartate aminotransferase, AST; alanine transaminase, ALT; gamma-glutamyl trans peptidase, γ -GT; lactate dehydrogenase, LDH), ferritin, transferrin, hormones (thyroid stimulating hormone, TSH; triiodothyronine, T3; thyroxine, T4) and a range of lipid parameters (total cholesterol, TChol; high density lipoprotein, HDL; low density lipoprotein, LDL; triglycerides, TG; ApoA1, apolipoprotein A1; ApoB, apolipoprotein B; lipoprotein A2, LPA2) were automatically analysed in the routine central laboratories of the Vienna General Hospital (Olympus 5400 clinical chemistry analysers, Beckman Coulter). All parameters were measured on the day of blood sampling.

UCB measurement (HPLC) in serum. For a detailed analysis of UCB (isomers), the method of HPLC was applied (after)⁵⁰, as had been used and published by our group^{4,51} and others⁵² previously. Briefly, fasting serum samples (stored light-protected in amber vials) were diluted in isocratic mobile phase (methanol, water, *n*-dioctylamine and acetic acid) and centrifuged. Supernatants were run on a chromatograph (Merck, Hitachi, LaChrom), equipped with a photodiode array detector (PDA, Shimadzu) and a Fortis C18 HPLC column ($4.6 \times 150 \text{ mm}$, $3 \mu\text{m}$), with a Phenomenex C18 HPLC guard column ($4 \times 3 \text{ mm}$). Sample preparation and analysis followed the previously published protocol⁴. Unconjugated bilirubin (Frontier Scientific Europe, Carnforth, Lancashire, UK) served as an external standard/quality control. As an internal standard, a reference serum sample was run in each analysis.

UGT1A1 Genotyping (-TA repeats in UGT1A1*28 promoter region). For UGT1A1 genotyping purposes, DNA was extracted from whole blood, using QIAAsymphony SP automated system with QIAAsymphony DSP DNA Midi Kit (QIAGEN), as instructed.

Analyses were performed as described elsewhere⁵³. Primers and probes were used as $10 \mu\text{M}$ working solutions. LightCycler FastStart DNA Master HybProbe Mix (Roche) was used on a LightCycler 480 Instrument II (Roche). Alleles were determined according to the melting curves obtained.

Anthropometric measurements. Standing height (subjects without shoes and in relaxed upright position) was measured with a commercial stadiometer, to the nearest 0.5 cm. Body mass (subjects barefooted and lightly dressed) was assessed to the nearest 0.1 kg, using digital scales. The body mass index (BMI) was calculated

following the equation $BMI = \text{body mass [kg]} / (\text{body height [m]}^2)^2$. To determine body composition, Bioelectric Impedance Analysis (BIA) was used, providing reliable data of body composition⁵⁴, and was performed in the mornings of the study days, using a BIA Analyser 2000-S (Data-Input GmbH, Darmstadt, Germany).

Lifestyle assessment of subjects. All participants were required to answer questions about their lifestyle, including (everyday) activity, exercise/training, drinking and eating habits. For this purpose, food frequency and lifestyle questionnaires were completed by each subject.

Indices of food intake were calculated for the reported weekly frequency of health food, snack food and red meat as well as alcohol consumption, and statistically analysed. “Health foods” included data on foods rich in vitamins, antioxidants, unsaturated fatty acids and fibres; “snack foods” referred to fatty and sugary energy-dense snacks; “red meat” included specifications on intake of red meat and meat products; “alcohol” referred to weekly alcoholic beverage consumption.

For details on weekly frequency of bodily activity, indices on overall activity, endurance exercise and resistance exercise were calculated. “Overall activity” included climbing stairs and walking; “endurance exercise” referred to frequency of at least 30 min bouts of endurance training; “resistance training” included reported frequency of resistance exercise (using own body weight and/or weights) per week.

Flow cytometric (FACS) analyses of pAMPK $\alpha 1/\alpha 2$, PgC1 alpha, pPpar alpha and $-\gamma$ in PBMCs. Active (phosphorylated) intracellular protein concentrations were measured in peripheral blood mononucleated cells (PBMCs). Cells were extracted from EDTA whole blood immediately after blood samplings. Density gradient centrifugation using separation tubes (LeucosepTM, Greiner bio one GmbH, Austria) was applied as instructed. Following isolation, cells were washed twice with ice-cold PBS. Cell count and viability were assessed using the trypan blue exclusion assay on an automated cell counter (CountessTM, Life Technologies). For short-term storage, cells were aliquoted in freezing medium (FBS + 10% DMSO) and gradually cooled (1 °C/min) to -80°C , using the CoolCellTM system (Biozym).

All FACS analyses were completed on a four-channel FACS CaliburTM flow cytometer (BD, Europe). Signal compensation (using CalibriteTM beads and FACS Comp software, BD) was successfully completed prior to each experimental run.

PBMCs were thawed at 37°C and washed twice with cold PBS (3000 g, 5 min). Cell count per test tube was adjusted to 250.000. After washing, cells were fixed (1% formalin, 10 min, RT), washed again, and permeabilised (70% ice-cold ethanol, 10 min, on ice). Following another washing step, cell pellets were suspended in staining-buffer (1% BSA, 0.02% Na-acide in PBS), and stained with respective antibodies (30 min, on ice; where applicable, the same is true for secondary antibody). All antibodies used had been titrated before use. Samples were run twice as independent duplicates, relative to respective negative/isotype controls. Antibodies were duplexed, so that cross-channel signal interference could be excluded. The antibody set-up used was as follows: anti-phospho-AMPK: rabbit anti-human monoclonal to AMPK $\alpha 1$ (phos-T183) and AMPK $\alpha 2$ (phos-T172), (ab133448, Abcam); secondary antibody to phospho-AMPK: goat anti-rabbit IgG H & L AlexaFluor 488, (ab150077, Abcam); anti-PgC1 α : rabbit anti-human polyclonal to PgC1 α , PE-labelled, (orb124814, Biorbyt); anti-phospho-Ppar α : rabbit anti-human polyclonal to Ppar α (phos-Ser12), FITC-labelled, (bs-4055R-FITC, Bioss); anti-phospho-Ppar γ : rabbit anti-human polyclonal to Ppar γ (phos-Ser112), AlexaFluor 647 labelled, (bs-3737R-A647, Bioss). Fluorescence signals (relative fluorescence units, rFU) were detected and recorded in the respective channels, and compared between study groups.

RNA extraction, cDNA synthesis and qPCR of AMPK1a gene expression. RNA was extracted from PBMCs using Qiagen RNeasy[®] Mini Kit, as instructed by the manufacturer, and using the QIAcube automated system. Total RNA concentration and quality were estimated using NanoDrop[®] ND-1000 (Thermo Scientific). The synthesis of cDNA from RNA was performed using the High-Capacity cDNA Reverse Transcription Kit with RNase Inhibitor 1000 Reactions (Applied Biosystems) on a Biomtra thermocycler. Concentration and quality of cDNA first strand was determined by NanoDrop 2000c spectrophotometer.

For qPCR 10ng cDNA samples were used on commercially available TaqMan assays (Life Technologies) as single-plex reactions. The TaqMan assay for PRKAA1 (assay Hs01562315_m1, FAM-MGB-labelled) gene was used according to manufacturer instructions. ACTB (assay Hs99999903_m1) and GAPDH (Hs99999905_m1) were used as endogenous controls.

The assays were performed using TaqMan Universal PCR master mix on a QuantStudioTM 6 Flex Real-Time PCR System (Thermo Fisher), on a 384-well block. All samples were run on the same 384-well plate in a single run, to avoid inter-plate variations. All samples were run in triplicate, and samples with a standard deviation higher than 0.5 Ct-units were excluded from further analyses. Relative quantification was performed using the RQ (Relative Quantification) feature on the Thermo Fisher Cloud qPCR analysis software with ACTB and GAPDH as endogenous controls and a pooled cDNA sample as a reference sample.

ELISA measurement of FGF-21 and Sirt-1 protein levels in serum. Fibroblast growth factor-21 (FGF-21) as well as Sirtuin-1 (Sirt-1) were measured in serum using respective ELISA kits (FGF-21 Human ELISA Kit, ab125966; Human SIRT1 ELISA Kit SimpleStep, ab171573; both Abcam), and following manufacturer instructions. Plate assays were run in a 96-well plate format, using a BMG FLUOstar OPTIMA microplate reader (BMG LABTECH GmbH), set to absorbance mode (450 nm). All samples and external standards were run in duplicate.

Statistical analyses. Statistics were completed using IBM SPSS 21 [IBM Corp. Released 2012. IBM SPSS Statistics for Windows, Version 21.0. Armonk, NY: IBM Corp.], as well as COVAIN toolbox for MATLAB

[MATLAB and Statistics Toolbox Release 2015a, The MathWorks, Inc., Natick, Massachusetts, United States.], for generating correlation heat maps.

Data distribution was checked using Kolmogorov-Smirnov (K-S test) and histograms. For comparison of means (for parametric data), ANOVA was used, for comparing medians or ranks (for non-parametric data), Mann-Whitney-U-test was selected. Data are summarized as is appropriate according to their respective distribution. For parametric data, means $\pm$ sd (standard deviation), for non-parametric variables, medians $\pm$ IQR (inter-quartile range) are presented. Bivariate correlations were modelled using Spearman correlation (Spearman's rho). Correlation coefficients (R) and p-values are presented. Regressions were calculated by applying the model of stepwise linear regression. Corrected R² coefficients and p-values are presented. For all statistical measures, the level of significance was set to be 5% ($p \leq 0.05$).

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Author Contributions

C.M. and M.W. planned, conceived and organised the study, were involved in recruiting subjects and completed a main part of experiments. C.M. completed the statistics and wrote the manuscript. C.K. analysed the key parameter UCB in all serum samples, and was in charge of processing blood samples. A.T. assisted in experiments (PCR, FACS, PBMC isolation), in data analysis and writing the manuscript. U.S. performed the Sirt-1 ELISA analyses. R.Z. completed the UGT1A1 genotyping and was involved in routine laboratory analyses. D.D. and R.M. completed the ethics, designed and conceived the study, oversaw the routine clinical measurements, and acted as medical consultants throughout the study. K.-H.W. provided financial support as well as designed and hosted the study. All authors critically revised the manuscript.

Additional Information

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