The Selection and Validation of Reference Genes for mRNA and microRNA Expression Studies in Human Liver Slices Using RT-qPCR

Tomáš Zárybnícký 1, Petra Matoušková 1, Martin Ambrož 1, Zdeněk Šubrt 2,3, Lenka Skálová 1 and Iva Boušová 1,*

1 Department of Biochemical Sciences, Charles University, Faculty of Pharmacy in Hradec Králové, 500 05 Hradec Králové, Czech Republic; zarybnto@faf.cuni.cz (T.Z.); matousp7@faf.cuni.cz (P.M.); ambrozm@faf.cuni.cz (M.A.); skaloval@faf.cuni.cz (L.S.)
2 Department of General Surgery, Third Faculty of Medicine and University Hospital Královoške Vinohrady, Charles University, 100 34 Prague, Czech Republic; subrt@email.cz
3 Department of Surgery, University Hospital Hradec Králové, 500 05 Hradec Králové, Czech Republic
* Correspondence: Iva.Bousova@faf.cuni.cz; Tel.: +420-495-067-406

Received: 29 August 2019; Accepted: 27 September 2019; Published: 28 September 2019

Abstract: The selection of a suitable combination of reference genes (RGs) for data normalization is a crucial step for obtaining reliable and reproducible results from transcriptional response analysis using a reverse transcription-quantitative polymerase chain reaction. This is especially so if a three-dimensional multicellular model prepared from liver tissues originating from biologically diverse human individuals is used. The mRNA and miRNA RGs stability were studied in thirty-five human liver tissue samples and twelve precision-cut human liver slices (PCLS) treated for 24 h with dimethyl sulfoxide (controls) and PCLS treated with β-naphthoflavone (10 µM) or rifampicin (10 µM) as cytochrome P450 (CYP) inducers. Validation of RGs was performed by an expression analysis of CYP3A4 and CYP1A2 on rifampicin and β-naphthoflavone induction, respectively. Regarding mRNA, the best combination of RGs for the controls was YWHAZ and B2M, while YWHAZ and ACTB were selected for the liver samples and treated PCLS. Stability of all candidate miRNA RGs was comparable or better than that of generally used short non-coding RNA U6. The best combination for the control PCLS was miR-16-5p and miR-152-3p, in contrast to the miR-16-5b and miR-23b-3p selected for the treated PCLS. Our results showed that the candidate RGs were rather stable, especially for miRNA in human PCLS.

Keywords: human liver; precision-cut liver slices; reference gene; RT-qPCR; miRNA; mRNA

1. Introduction

Precision-cut liver slices (PCLS) are a three-dimensional multicellular model with preserved tissue architecture as well as cell–cell and cell–matrix interactions that maintain and mimic liver in vivo functions. [1,2] The applicability of PCLS includes toxicity, inflammation and oxidative stress [3–6], and fibrosis as well as expression changes upon short-term or prolonged incubation of PCLS [7–12].

Choosing suitable reference genes (RGs) for the normalization of gene expression is always a dilemma. Reverse transcription-quantitative polymerase chain reaction (RT-qPCR) is a very potent tool with high sensitivity, accuracy, specificity, reproducibility, and relatively low cost in comparison to microarrays or high-throughput sequencing [13]. Nevertheless, RT-qPCR can be influenced by factors such as RNA integrity, genomic DNA contamination, reverse transcription reaction efficiency, and complementary DNA (cDNA) quality [14]. There have been no studies to properly validate the stability of RGs despite the spreading usage of PCLS, although PCLS are a useful model to study liver injury as...
As not all liver samples were suitable for PCLS preparation, small pieces of twenty-three liver samples were only placed into TRI Reagent solution (Biotech, Prague, Czech Republic) and stored in a freezer at –80 °C until RNA isolation. Twelve liver samples were used for PCLS preparation according to the below mentioned protocol.
Table 1. Summary of human samples.

| Human Sample | Gender (Age) | Reason of Surgery | Long-Term Pharmacotherapy | PCLS Preparation |
|--------------|--------------|-------------------|---------------------------|------------------|
| L1           | Female (37)  | CRC \(^1\)        | Insulin                   | No               |
| L3           | Male (58)    | CCC \(^2\)        | Allopurinol, felodipine, ramipril, indapamide, atorvastatin, cilostazol, pregabalin | No               |
| L4           | Male (35)    | Adenoma           | Mesalazine, omeprazole, escitalopram, budesonide | No               |
| L5           | Male (63)    | CRC               | Ramipril, atorvastatin, metformin, allopurinol | Yes              |
| L6           | Male (69)    | CRC               | Hydrochlorothiazide       | Yes              |
| L7           | Male (69)    | CRC               | Nifedipine, acetylsalicylic acid | Yes              |
| L8           | Female (69)  | CRC               | Verapamil, trandolapril   | No               |
| L9           | Male (58)    | CRC               | Betaxolol                 | Yes              |
| L10          | Female (66)  | HCC \(^3\)       | Pantoprazole              | No               |
| L11          | Female (57)  | CRC, liver metastases | None                     | Yes              |
| L12          | Female (73)  | CRC               | Lercanidipine, furosemide, perindopril, nadroprin | No               |
| L13          | Female (67)  | CCC               | Nadroprin                 | No               |
| L14          | Female (45)  | BFNH \(^4\)      | None                      | Yes              |
| L15          | Female (69)  | CRC               | Nebivolol, simvastatin, digoxin, irbesartan, hydrochlorothiazide, nadroprin | No               |
| L16          | Female (59)  | CRC               | None                      | Yes              |
| L17          | Male (39)    | CRC               | None                      | No               |
| L18          | Male (83)    | HCC               | Lacidipine, solifenacin, tamsulosin | No               |
| L19          | Female (65)  | CRC               | Amlodipine                | No               |
| L20          | Female (84)  | Abscess           | None                      | No               |
| L21          | Male (34)    | Jejunal adenocarcinoma | None                     | No               |
| L22          | Male (84)    | CRC               | Telmisartan, nifedipine, formoterol | No               |
| L23          | Male (83)    | CRC               | Furosemide, amiodipine, acetylsalicylic acid, telmisartan, salmeterol, fluticasone | No               |
| L24          | Male (77)    | HCC               | Metoprolol, felodipine, ramipril, metformin, glipizide, finasteride, warfarin, nadroprin | No               |
| L25          | Male (70)    | CRC               | Perindopril, betaxolol, metformin, atorvastatin, allopurinol, insulin | No               |
| L26          | Male (72)    | CRC               | None                      | No               |
| L27          | Male (70)    | CRC               | Insulin, atorvastatin     | No               |
| L28          | Female (26)  | BFNH              | None                      | Yes              |
| L29          | Male (59)    | Renal cell carcinoma, liver metastases | None                     | No               |
| L30          | Female (81)  | CRC, liver metastases | Hydrochlorothiazide, betaxolol, acetylsalicylic acid, zolpidem, insulin | Yes              |
| L31          | Female (72)  | CCC               | Lercanidipine, irbesartan, hydrochlorothiazide, lansoprazole, levothyroxine, acetylsalicylic acid, fenofibrate, nebivolol | No               |
| L32          | Male (62)    | CRC               | Ramipril, felodipine, metoprolol, rosuvastatin, acetylsalicylic acid | No               |
| L33          | Male (62)    | CRC, liver metastases | Tamsulosin, metoprolol, omeprazole | No               |
| L34          | Female (78)  | CCC               | Simvastatin, bisoprolol, furosemide, ramipril, enoxaparin, zolpidem | Yes              |
| L35          | Male (50)    | Neuroendocrine tumor, liver metastases | Insulin                   | Yes              |
| L36          | Male (59)    | CCC               | None                      | Yes              |

\(^{1}\) CRC, colorectal carcinoma; \(^{2}\) CCC, cholangiocellular carcinoma; \(^{3}\) HCC, hepatocellular carcinoma; \(^{4}\) BFNH, benign focal nodular hyperplasia.
2.4. Preparation of Precision-Cut Liver Slice and Experimental Treatment

The PCLS were prepared according to Zárybnický et al. [4]. Briefly, cylindric cores were cut out of the tissue. A Krumdieck tissue slicer MD4000 (Alabama Research and Development, Munford, AL, USA) filled with a supplemented ice-cold Krebs–Henseleit buffer saturated with carbogen (95% O₂, 5% CO₂) was used to prepare the PCLS (thickness: 150 µm, diameter: 8 mm). The incubation of PCLS proceeded individually in 12-well plates filled with supplemented Williams’ Medium E with GlutaMAX (25 mM glucose and 50 µg/mL gentamycin; ThermoFisher, Waltham, MA, USA) at 37 °C under continuous supply of 85% O₂/5% CO₂ (Binder CB 60 incubator, Tuttingen, Germany) with continuous shaking (90 times/min). All experiments were performed in triplicate. The slices were treated with 0.1% DMSO (controls), 10 µM rifampicin (RIF, a strong CYP3A4 inducer), or 10 µM β-naphthoflavone (BNF, a CYP1A1/2 inducer) [19].

2.5. Viability

Viability of the PCLS was measured in triplicate at multiple time points: −1 (before pre-incubation), 0 (after pre-incubation), 4, 8, 12, 18, and 24 h. The levels of adenosine triphosphate (ATP) and leakage of lactate dehydrogenase (LDH) were used as viability markers. The ATP levels were determined by the method by Hadi et al. [20], according to the protocol described in our previous report [4]. The LDH leakage into the culture medium, expressed as % of total LDH activity, was determined as a marker of plasma membrane damage. For the analysis, all of the culture medium was taken up and kept in −80 °C until use. After thawing, the medium was centrifuged at 2000 rpm for 5 min at 4 °C to remove cellular debris. To determine the total LDH activity, three slices were incubated for 24 h, and afterward, each slice was homogenized in its culture medium by ribbed shaft pestles, with these homogenates processed in the same way as the tested culture mediums. LDH activity was determined using the LDH cytotoxicity assay (Roche Diagnostics, Mannheim, Germany). The absorbance of technical duplicates was measured using the plate reader Infinite M200 (Tecan Group, Mannedorf, Switzerland).

2.6. Tissue RNA Extraction

The PCLS were collected after 4, 8, 12, 18, and 24 h. All treatments were performed in triplicate and every slice was put separately into 500 µL of TRI Reagent and stored at −80 °C until use for at most up to two months. Total RNA isolation from each slice was performed according to the manufacturer’s instructions (TRI Reagent®, Biotech, Prague, Czech Republic). The sample homogenization was carried out using a single metal bead in a 2 mL Eppendorf tube using a microhomogenizer. Forty µL of diethyl pyrocarbonate (DEPC)-treated water (0.01% DEPC in HPLC water, autoclaved) was used to dissolve the purified RNA, which was then stored at −80 °C. RNA yields and purity were evaluated by measuring the absorbance at 260 and 280 nm using a NanoDrop ND-1000 UV–Vis Spectrophotometer (Thermo Fisher Scientific, Prague, Czech Republic), where all samples had a ratio of absorbance at 260 nm and 280 nm higher than 1.8. RNA integrity numbers (RIN) were detected by an Agilent 2000 analyzer (Agilent, Santa Clara, CA, USA), with the RIN of all samples greater than 6.0 (Table S1). Furthermore, DNase treatment was performed to avoid genomic DNA contamination. Details of this procedure are described in Table S2. The DNase I treated RNA was stored at −80 °C until further analyses.

2.7. cDNA Synthesis

Details of cDNA synthesis for mRNA, miRNA, and U6 are summarized in Table S2. For mRNA and miRNA, the obtained cDNAs were diluted 1:6 and 1:9 with DEPC water, respectively. All cDNAs were stored at −20 °C until the qPCR assay.
2.8. Primer Design, Quantitative Real-Time PCR

Descriptions of selected candidate RGs for normalization of mRNA and miRNA as well as genes of interest are mentioned in Table 2. For mRNA normalization, the primers were either described in previous reports or designed manually using a Primer3 software [21] targeting sequence outside the hairpin structures calculated for 60 °C using mFold [22], with the specificity checked using NCBI primer blast. The reverse primer used for miRNA normalization was universal (the sequence inserted by the Stem-Loop primer) [23,24] and the forward primers were designed manually and controlled using OligoCalc [25]. Synthesis of all primers was carried out by Generi Biotech (Hradec Králové, Czech Republic). The primer sequences and other information are also listed in Table 2. Melting curve plots are presented in Supplementary Materials (Figure S1).

RT-qPCR was performed following the minimum information for publication of quantitative real-time PCR experiments (MIQE) guidelines [17] and all experimental details are stated in Table S2.

2.9. Data Analysis

Since the PCLSs from each human were processed separately, each analysis was performed individually. Two types of software were used to assess the stability of the candidate RGs, and the analyses were based on untransformed Ct values. RefFinder is a freely available webtool, the main advantage of which is its ability to integrate four computational algorithms (geNorm [13], BestKeeper [26], NormFinder [27], and Comparative ∆CT method [28]) [29]. RefFinder ranking of the genes is based on the geometric mean of placings from each of the four above-mentioned programs [29]. Although the stability ranking of the candidate genes is important, it does not provide information regarding how many reference genes are necessary for normalization. Therefore, we additionally used the freely available Microsoft Excel geNorm version (inputting the transformed Ct values into relative quantities) to estimate the optimal number of RGs that should be used for the intended experiment via the calculation of pairwise variation values [13] (the geNorm output is not included in RefFinder).

Relative expression levels of the target genes were calculated as the fold change from the triplicates in each group using the $2^{-\Delta\Delta CT}$ method [30]. Results were expressed as the mean ± SD. Comparisons among the control and treated groups were performed using a two-way analysis of variance (ANOVA), using GraphPad Prism 7 (GraphPad Software, La Jolla, CA, USA). Differences were considered significant at $p < 0.05$. 
Table 2. Description of selected candidate reference genes and genes of interest.

| Gene Symbol | Gene Name | GeneBank or miRBase Accession Number | Gene Function | Primer Sequences 5'→3' | Tm\(^1\) (°C) | E\(^2\) (%) |
|-------------|-----------|-------------------------------------|---------------|-------------------------|--------------|------------|
| **Candidate reference genes for mRNA normalization** | | | | | | |
| ACTB | Actin beta | NM_001101.4 | Structural protein of cytoskeleton | F: TCCCTGGAGAAGACCTACGAG<br>R: CAGGAAGGAGGCTGGAAGAG | 86.5 | 102 |
| B2M | Beta-2-microglobulin | NM_004048.2 | Beta-chain of major histocompatibility complex | F: TCTCGTCTCATGGTTTGATGC<br>R: TCCTGCTCCCCACCTCCTAAG | 83 | 99 |
| GAPDH | Glyceraldehyde-3-phosphate dehydrogenase | NM_002046 | Enzyme of glycolysis pathway | F: GAGGCATGTGATAGCTGAG<br>R: TGTCACCGAGATATCACAAAGA<br>R: TCTCAATTCACAAAGATGTTCGCT | 86 | 101 |
| HPRT1 | Hypoxanthine phosphoribosyltransferase 1 | NM_000194.2 | Metabolism of purines | F: TGAGGATGCTGATAGCTGAG<br>R: TCTCAATTCACAAAGATGTTCGCT | 82 | 101 |
| SDHA | Succinate dehydrogenase complex, subunit A | NM_004168.3 | Critical function in mitochondrial respiratory chain | F: TGGGAACAAGAGGGCATCTG<br>R: ACCACACTGCTCATTACATG | 79.5 | 99 |
| YWHAZ | Tyrosine 3-monooxygenase/tryptophan 5-monooxygenase activation protein zeta | NM_003406.3 | Important protein for many signal transduction pathways | F: TGTACCCCAATGCTTCAAG<br>R: GCCAACGATCCGATACG | 77.5 | 102 |

| **Candidate reference genes for miRNA normalization** | | | | | | |
| miR-16-5p | MicroRNA 16 (5p) | MIMAT0000069 | Regulation of apoptosis | F: GTCTCCTCTGTGCCGCCAGAG<br>GATGCCAACCAGCAAGCCCGC AA<br>F: ACACGCTACGACGGTAAAT | 79 | 102 |
| miR-23b-3p | MicroRNA 23b (3p) | MIMAT0000418 | Associated with cell proliferation, invasion, and apoptosis | F: ATCTGTATCAGTTCCAGG<br>TGCACCAGCAAGCCCGC AA<br>F: ACACGCTACGACGGTAAAT | 77.5 | 109 |
| miR-93-5p | MicroRNA 93 (5p) | MIMAT0000093 | OncomiR, plays an essential role in tumorigenesis and progression of various carcinomas | F: ATCTGTATCAGTTCCAGG<br>TGCACCAGCAAGCCCGC AA<br>F: ACACGCTACGACGGTAAAT | 78.5 | 105 |
| miR-152-3p | MicroRNA 152 (3p) | MIMAT000000438 | Regulates hepatic glycogenesis, tumor suppressor | F: ATCTGTATCAGTTCCAGG<br>TGCACCAGCAAGCCCGC AA<br>F: ACACGCTACGACGGTAAAT | 78.5 | 100 |
| U6 | Small nuclear RNA U6 | NR_003027 | RNA splicing | F: GCTCGTTTCCGACAGCA<br>R: GAGGTATTCGACCAGAGGA<br>F: GCTCGTTTCCGACAGCA<br>R: GAGGTATTCGACCAGAGGA | 80.5 | 99 |

| **Genes of interest** | | | | | | |
| CYP1A2 | Cytochrome P450 family 1 subfamily A member 2 | NM_000761 | Phase I biotransformation | F: CCTCCTGAGACACTGAG<br>R: GCCACCTGACCAACTCAG<br>F: CCTCCTGAGACACTGAG<br>R: GCCACCTGACCAACTCAG | 85.5 | 101 |
| CYP3A4 | Cytochrome P450 family 3 subfamily A member 4 | NM_017460.5 | Phase I biotransformation | F: CCTCCTGAGACACTGAG<br>R: GCCACCTGACCAACTCAG<br>F: CCTCCTGAGACACTGAG<br>R: GCCACCTGACCAACTCAG | 82.5 | 99 |
| miR-27a-3p | MicroRNA 27a (3p) | MIMAT000084 | Direct regulation of CYP3A4 | F: CCTCCTGAGACACTGAG<br>R: GCCACCTGACCAACTCAG<br>F: CCTCCTGAGACACTGAG<br>R: GCCACCTGACCAACTCAG | 80.5 | 106 |
| miR-203a-3p | MicroRNA 203a (3p) | MIMAT0000264 | Indirect regulation of CYP1A2 via PXR receptor | F: CCTCCTGAGACACTGAG<br>R: GCCACCTGACCAACTCAG<br>F: CCTCCTGAGACACTGAG<br>R: GCCACCTGACCAACTCAG | 78.5 | 105 |

\(^1\) Tm, melting temperature; \(^2\) E, assays efficiency; \(^3\) F, forward primer; \(^4\) R, reverse primer; \(^5\) RT, reverse transcription primer.
3. Results

All liver samples included in this study were selected to be as healthy as possible even if patients were being treated under different clinical conditions. Based on the steatosis (score of 0–1) and fibrosis (score of zero) scoring performed by a pathologist, all the liver biopsies showed no or mild signs of liver disease. As biochemical markers of liver functions, plasma levels of BIL, ALT, AST, GMT, and ALP were assessed (Table S3). All of the assessed biochemical parameters were within physiological ranges in thirteen samples, while two samples (L6 and L23) exerted an elevated BIL level, probably as a result of intrahepatic obstruction due to a malignancy. ALP only showed an increase in four samples, with GMT only elevated in twelve samples; a combination of ALP-GMT and AST-GMT was elevated in two samples and one sample, respectively. Moreover, plasma levels of ALT, AST, GMT, and ALP were increased in L21 and L37 with an AST/ALT ratio of 0.62 and 0.59, respectively. Routinely tested liver biomarkers (e.g., ALT, AST, GMT, BIL) are altered in different malignities and hepatic metastases as well as in patients with primary or secondary malignant involvement of the hepatic system, all of which represent significant risk predictors for all-cause mortality [31].

3.1. Analysis of Candidate RGs Expression Stability for mRNA Normalization and Their Validation in Human Liver Tissue

With regard to the information obtained from literature concerning RGs in liver/PCLS, a set of six candidate RGs, namely beta-actin (ACTB), beta-2-microglobulin (B2M), glyceraldehyde 3-phosphate dehydrogenase (GAPDH), hypoxanthine phosphoribosyltransferase 1 (HPRT1), subunit A of succinate dehydrogenase complex (SDHA), and tyrosine 3-monooxygenase/tryptophan 5-monooxygenase activation protein zeta (YWHAZ) was tested in human liver samples collected from patients undergoing partial hepatectomy [3,7,9,10,32,33]. The raw quantification cycle (Cq) values were obtained for all thirty-five human liver tissue samples (Figure 1A). In this study, HPRT1 was the candidate RG with the lowest expression (mean Cq = 27.76, SD = 0.50) as well as the lowest variability (coefficient of variation, CV = 1.80%), while ACTB was the most abundantly expressed gene (mean Cq = 20.36, SD = 1.34). The highest variability was found for GAPDH with CV = 6.69%.

The ranking of the expression stability of RGs was performed using RefFinder [29], which integrates four computational algorithms to compare and rank the candidate RGs. Based on the rankings from each program, RefFinder assigns a corresponding weight to an individual gene and calculates the geometric mean of their weights for the overall final ranking. Moreover, the ranking of RGs based on their expression stability measure (M), which represents the average pairwise variation of a certain gene with all other tested reference genes [13], was performed using geNorm. The most stably expressed RGs in the human liver samples were YWHAZ and ACTB, with an M value of 0.742 (Figure 1B). The raw data from RefFinder are presented in Table S4.

In addition, geNorm enables the calculation of the pairwise variation (Vn/Vn+1) between two sequential candidate RGs to define the optimal number of RGs that should be used. Vandesompele et al. (2002) defined a pairwise variation of 0.15 as the cut-off value, which indicates that the inclusion of an additional RG is needless [13]. In our analysis, the V2/3 value was 0.017, suggesting two RGs as optimal (Figure 1C).

The use of different RGs to calculate the relative expression data of target genes could significantly affect the final normalized results. To show the impact of different RGs on the result of the practical experiment, the relative expression patterns of CYP1A2 and CYP3A4 were analyzed, with differences shown for six selected liver samples (Figure 1D,E, rest in Table S5). Different normalizations were applied: (1) the most stable gene (YWHAZ); (2) the second most stable gene (ACTB); (3) the geometric mean of YWHAZ and ACTB; (4) the least stable gene (B2M); and (5) the geometric mean of two of the least stable genes (B2M/HPRT1). In all liver samples, the expression levels of both CYPs were comparable when the three most stable normalizations (1–3) were applied. Moreover, the way of normalization did not influence the expression level of the target genes in some of the liver samples (e.g., L10 and L23). However, when CYP1A2 and CYP3A4 were normalized using normalizations 4–5,
the expression levels of both CYPs increased several times when compared to normalizations 1–3 in many of the liver samples (e.g., L17, L19, L28, and L36).

3.2. Analysis of Candidate RGs Expression Stability for mRNA Normalization in Human PCLS

Subsequently, results obtained with human liver tissue were compared to the human PCLS. Analysis of candidate RGs stability in human PCLS was performed in two ways. In the first part, PCLS obtained from the liver of twelve patients were incubated in the medium containing DMSO (controls) or the CYP inducers BNF and RIF (treated groups) for 24 h, following which the expression of RGs was assessed. In the second part, the mRNA expression of the RGs was monitored at five time points within a 24 h incubation of PCLS with DMSO or CYP inducers. Such an experiment is demanding with regard to the amount of the already rare tissue, therefore, only liver samples obtained from three individuals were used for this experiment.

The first and crucial step in this experiment was to ensure optimal culture conditions for the PCLS, otherwise the expression rate of the tested genes could be unpredictably influenced. Levels of
ATP and leakage of LDH were used as viability markers. ATP levels increased after pre-incubation, a finding which could be explained by the restoration of mitochondrial functions after the ischemic period during surgery. ATP levels remained rather stable for the rest of the experiment. At each time-point, LDH leakage usually did not exceed 10% of total LDH activity, which denotes that the plasma membranes of the liver cells were not significantly damaged (Figure 2).

![Figure 2. Viability results of human precision-cut liver slices (PCLS) during 24-h incubation. (A) adenosine triphosphate content; (B) lactate dehydrogenase leakage. Results are presented as mean ± SD (n = 3).](image)

3.3. Reference Genes Ranking According to Their Expression Stability

After performing RT-qPCR, the Cq values from the controls and treated groups were analyzed separately. Slices from each human sample were analyzed individually, since they were also processed and measured separately; in addition, potential biological variation can be shown this way. RefFinder was used to create an overall comprehensive ranking of the most stable RGs for our defined groups, with raw data shown in Table S6 and Table S7. Moreover, the V2/3 value was in all cases lower than 0.01, suggesting two RGs as optimal for normalization (Figures S2 and S3).

The best combination of RGs was selected based on the overall rankings of RGs in RefFinder for individual human samples. Optimally, the gene stability should be equal among the samples, however, biological variation plays an unpredictable role, and therefore we recommend a combination of genes for which the overall rankings were lowest among human samples. According to RefFinder, the most stable genes for 24 h incubated PCLS in the controls were YWHAZ and B2M, while ACTB and YWHAZ were the most stable genes in the treated groups (Figure 3A,C). It is important to emphasize that differences in stability of our tested RGs were small (rankings in Figure 3B,D).

For the RGs' expression stability within 24 h incubation, all the tested RGs showed a stable expression during incubation. In the controls, the stability order (shown in descending stability manner) was as follows: YWHAZ > ACTB > B2M > HPRT1 = SDHA > GAPDH, while the stability order in the treated groups was YWHAZ > B2M > ACTB > HPRT1 = SDHA = GAPDH (Figure 4A,B). GeNorm analysis showed that a combination of two RGs was sufficient for normalization (V2/3 < 0.01 for all PCLS). In summary, YWHAZ + ACTB seems to be a suitable potential combination of RGs for induction studies or projects studying changes in human PCLS during incubation. Despite the finding that B2M ranks higher among treated groups than does ACTB, in our liver samples, B2M was the least stable gene and therefore we would not recommend its use as a RG. The raw RefFinder data and results of geNorm analysis are presented in Tables S8 and S9 and Figures S4–S6.
when using normalization to YWHAZ, CYP3A4, and Genes expression was similar in samples L28 and L37. Interestingly, there was a significant disparity between incubation, di gene’s mRNA induction reached its maximum, which was found to be reached between 18 to 24 h of respectively (Figure 4D,E). These results also enabled us to identify the interval when the target BNF and RIF on the expressions of CYP1A2 incubation period.

Nevertheless, Figure 4 shows how much the rate and trend of target gene expression can be influenced by the RG selected in experiments using PCLS. When using ACTB as a RG, the expression of CYP3A4 increases 6–10 times within the first 18 h of incubation and then starts to decrease, while when using normalization to YWHAZ, the CYP3A4 expression profile rises throughout the whole incubation period.

In order to evaluate the chosen RGs, we decided to compare the effects of the model CYP inducers BNF and RIF on the expressions of CYP1A2 and CYP3A4 using normalization to YWHAZ/B2M, respectively (Figure 4D,E). These results also enabled us to identify the interval when the target gene’s mRNA induction reached its maximum, which was found to be reached between 18 to 24 h of incubation, differing among the individual human samples. The tendency in CYP1A2 and CYP3A4 expression was similar in samples L28 and L37. Interestingly, there was a significant disparity between CYP3A4 and CYP1A2 in terms of reaching their maximum in sample L30.

3.4. Validation of Reference Genes for mRNA Normalization

The selection of RGs could have played an important role in the calculation and interpretation of our final results, especially for the time dependent findings. To show the impact of different RGs on the results of a practical experiment, the fold change of CYP3A4 in sample L28 was analyzed (Figure 4C). Careful selection of RGs prior to our study showed that all of the tested candidate RGs were satisfactory and could be used for normalization since they are quite stable over time. Nevertheless, Figure 4 shows how much the rate and trend of target gene expression can be influenced by the RG selected in experiments using PCLS. When using ACTB, B2M, GAPDH, HPRT, or SDHA as a RG, the expression of CYP3A4 increases 6–10 times within the first 18 h of incubation and then starts to decrease, while when using normalization to YWHAZ, the CYP3A4 expression profile rises throughout the whole incubation period.
Figure 4. RefFinder ranking of RGs with individual values in PCLS within 24-h incubation in the control (A) and treated group (B); (C) Effects of different reference genes on the normalization of CYP3A4 expression in sample L28. Results are presented as mean ± SD (n = 3); (D, E) Relative mRNA expression of CYP3A4 (D) and CYP1A2 (E) under the effect of rifampicin (RIF) 10 µM and β-naphthoflavone (BNF) 10 µM, normalized to geometric mean of YWHAZ and B2M. Results are presented as mean ± SD (n = 3).

3.5. Analysis of Expression Stability and Validation of RGs for miRNA Normalization in Human PCLS

The last part of this study focused on the selection and validation of suitable RGs for miRNA expression studies. The PCLS for miRNA expression studies were obtained from the livers of three individuals, and miRNA expression was studied at five time points within 24 h incubation. Following the information obtained from the literature, a set of five candidate miRNA RGs, namely U6, miR-16-5p, miR-23b-3p, miR-93-5p, and miR-152-3p, was tested.

According to the RefFinder analysis, the stability order (shown in descending stability manner) was as follows: miR-16-5p > miR-152-3p > U6 = miR-93-5p > miR-23b-3p. In the treated group, the gene stability decreased in the order miR-16-5p > miR-23b-3p > miR-152-3p > U6 = miR-93-5p (Figure 5A,B). Therefore, the most stable genes in the control group were miR-16-5p and miR-152-3p,
while in the treated group, the most stable genes were miR-16-5p and miR-23b-3p. GeNorm analysis showed that a combination of two RGs was sufficient for miRNA normalization (V2/3 < 0.15 for all PCLS). Taken together, the combinations of miR-16-5p + miR-152-3p/miR-23b-3p showed promising results in the tested human samples and these combinations seem to be suitable potential RGs for induction studies or projects studying changes in human PCLS during incubation. The observed stability differences for the tested miRNAs were almost negligible. Raw RefFinder data and the results of the geNorm analysis are presented in Tables S10 and S11 and Figures S6 and S7.

Figure 5. RefFinder ranking of miRNA RGs with individual values in PCLS within 24-h incubation in the control (A) and treated group (B). Relative miR-27a-3p and miR-203a-3p expression under the effect of RIF 10 µM (C) and BNF 10 µM (D), respectively, normalized to a geometric mean of miR-16-5p and miR-23b-3p. Results are presented as mean ± SD (n = 3).

The relative expression of miR-27a-3p and miR-203a-3p in the PCLS treated with RIF and BNF are presented in Figure 5C,D, respectively. Both miR-27a-3p and miR-203a-3p revealed inter-individual variability rather than the statistically significant changes caused by the inducers.

4. Discussion

PCLS are a model with increasing use among laboratories. Nevertheless, heterogenic handling and protocols can influence the viability and obtained results from PCLS. The issue of protocols/handling and incubation systems among laboratories was summarized by Granitzny et al. [34] and we hope that the present manuscript will provide an impetus to discuss the selection of RGs for RT-qPCR. Although the use of microarrays and next generation sequencing is becoming more widespread, these techniques have their own limitations (e.g., data normalization), and it is often recommended to validate their results through RT-qPCR. Normalization of RT-qPCR data is performed in order to eliminate sampling differences (e.g., RNA quality and quantity) as well as to observe real gene-specific variations, which
should be none or minimal for a suitable internal RG. Therefore, determining suitable endogenous RGs is a crucial step in the analysis of gene expressions, as the employment of an unstable gene for data normalization could distort the obtained results and thus cause the wrong conclusions to be drawn.

After surveying the literature regarding mRNA normalization in PCLS and not deliberately considering the differences among species, we found that various single or combined RGs have been used, for example, ACTB [3,7], B2M [3], GAPDH [3,7,9,10], HPRT [3,7,35], RPLPO [3], GUSB [35], α-SMA [11], and in one case, the particular RG was not mentioned [36]. Four of the most frequently used RGs (ACTB, B2M, GAPDH, HPRT) were used in this study, and two more (SDHA, YWHAZ) were added into the tested pool of genes based on their status as RGs in liver tissue [32,33].

In the first phase of study, the stability of the selected RGs in thirty-five human liver tissues was verified. Based on the comprehensive ranking of individual liver samples in Reffinder, the most stably expressed RGs were YWHAZ and ACTB. According to the geNorm analysis, a combination of two RGs is sufficient for mRNA normalization in human liver samples. As shown in Figure 1D,E, the selection of RG can markedly influence the expression levels of target genes, in our case, CYP1A2 and CYP3A4.

Next, the stability order of the selected genes was checked in the larger pool of PCLS (twelve samples) only after 24 h of cultivation. In the second phase, it was verified in three liver samples that the tested RGs were capable of stable expression within 24-h cultivation in the model of PCLS. Both studies were performed under the same conditions and analyzed in the same way (controls vs. treated group). However, it is difficult to draw a definite conclusion from this experiment, as the sample size was too small. Ideally, a larger set of human samples would be necessary for a detailed study of the time profile of the candidate RGs’ expression within incubation. Unfortunately, the measurement of continuous expression is highly demanding with regard to the amount of the already extremely precious tissue.

miRNAs are short non-coding RNAs with 20–22 bases mediating post-transcriptional gene regulation. miRNAs usually bind to the 3′UTR of specific mRNAs to suppress their translation or accelerate degradation [37,38]. It is worth mentioning that a single miRNA can bind to several transcripts, and a single transcript can possess binding sites for multiple miRNAs [39]. The problem of miRNA normalization is the limited pool of RGs for many models, tissues, and body fluids. However, finding even a single stable miRNA under certain experimental conditions can be tricky, and therefore other non-coding RNAs such as small nuclear RNA U6 or small nucleolar RNAs (e.g., 204, 234, U24, or U26) are used quite often as RGs [37,38]. Nevertheless, the biochemical character of these non-coding RNAs does not reflect that of miRNAs, and the efficiency of extraction and reverse transcription might also differ from that of miRNAs [38,40,41]. In terms of miRNAs, we only found three publications regarding RT-qPCR data normalization. While two of these studies used only U6 for normalization [5,6], the third one used miR-93-5p [4] in PCLS.

In our experiment, miR-16-5p was among the most stable tested genes, while in a previous study of ours with human PCLS treated by three different hepatotoxicants (i.e., acetaminophen, R-pulegone and R-menthofuran), miR-93-5p was more stable than U6 and miR-16-5p. Having previous experience with miR-16-5p and miR-93-5p as potential RGs, the pool of candidates was enlarged based on the study of Lamba et al., who analyzed multiple human liver samples in search of RG candidates [42]. Based on the analysis of these researchers, miR-23b and miR-152 were selected. We verified several miRNA RGs whose stability was comparable or better than U6, giving us a better chance to avoid using short non-coding RNAs that were not miRNAs for data normalization. We believe that the tested miR-16-5p, miR-23b-3p, miR-93-5p, and miR-152-3p could also be used in validations with other liver models.

When designing an experiment, the selection of the time point of interest is very important. Unfortunately, human tissues for experimentation are scarce and difficult to obtain, and it is not always possible to perform a similar experiment with positive/negative controls and compounds of interest at multiple concentrations and at multiple time points. In optimal conditions, a single time point for the whole experiment is strongly preferred. Based on the BNF and RIF results (Figure 4), we decided...
that the time point of 24 h of incubation was sufficient to decide how strong a CYP inducer/tested compound is. Nevertheless, multiple timepoints would be needed to find the maximum expression if the goal is to compare inter-individual variation.

The multicellularity of PCLS in terms of gene expression represents both an advantage and risk, since this factor can influence the mRNA and miRNA expression profiles. Gene expression is regulated by different mechanisms, one of these being negative regulation by microRNAs. The expression of CYPs is also partly regulated by miRNAs [43]. The expression of CYP3A4 is directly influenced by miR-27a-3p, while miR-203a-3p indirectly regulates CYP1A2 via the pregnane X receptor [44]. The relative expression of these miRNAs (Figure 5) shows inter-individual variability with no trend similarities and no statistically significant changes. In our search for publications describing the experimental results regarding the influence of BNF or RIF on miRNA profiles in the liver, we did not find any information about BNF, although two papers were found about the influence of 10 µM RIF on miRNA expression in primary human hepatocytes using microarrays. The results of these two articles presented a 3-fold downregulation of miR-27a after 24 h [45] and a 1.82-fold upregulation after 48 h [46], respectively. However, the miRNA expression in the first case was analyzed by differential sequencing, and in the second case by RT-qPCR normalized only to U6. On the other hand, the results from our experiments (Figure 5) displayed interindividual dependent variation rather than time- or induction-influenced variation. In the cases of CYP3A4 and CYP1A2, the multicellularity did not seem to pose a problem. The multifold induction of CYP3A4 by rifampicin and CYP1A2 by BNF were well visible, showing interindividual differences in the rate of induction as well as in the time it took these genes to reach the peak of induction, with a peak interval of between 18 to 24 h for both genes (Figure 4).

RT-qPCR has become a routine technique, allowing the high-throughput analysis of RNA expression over a large range and at relatively low cost. Nevertheless, normalization remains a problem and it is a target of frequent criticism [47]. It would be optimal if a comparison of a selection of RGs among various laboratories and research groups could be undertaken, not to mention a comparison between the stability of RGs in human liver slices and PCLS from other species. Human tissues are especially problematic due to their variability and inter-individual differences. While HPRT1 was the least stable in samples L28 and L30, it was the most stable gene in L38.

5. Conclusions

In summary, a set of potential RGs for mRNA was tested in thirty-five human liver samples using RT-qPCR, with the combination of YWHAZ/ACTB identified in geNorm and RefFinder as the most suitable for gene normalization. Subsequently, two sets of potential RGs for mRNA and miRNA were tested in human PCLS at multiple time points during a 24 h incubation using RT-qPCR. A combination of two RGs was identified in geNorm as sufficient for gene normalization. The most stable RGs, based on the comprehensive ranking of individual liver samples in RefFinder, were determined to be YWHAZ/B2M for the control slices and YWHAZ/ACTB for the treated slices. However, the difference in the stability results for candidate RGs was small, and all of them could be used for normalization. Nevertheless, B2M was the least stable candidate RG in the human liver samples and therefore we would not recommend its use for data normalization in PCLS. In the case of miRNA, the most stable genes in the control and treated PCLS were miR-16-5p/miR-152-3p and miR-16-5p/miR-23b-3p, respectively, all of which were equally or more stable than the commonly used U6. The genes were found to be rather stable, especially for miRNA in the tested human PCLS. The stability of RGs must always be tested due to many variables involved, for example, variabilities in liver donor pharmacotherapy, donor diet, pathologies, and age as well as experimental conditions and handling.

Supplementary Materials: The following are available online at http://www.mdpi.com/2073-4425/10/10/763/s1,
Figure S1: Melting curve plots; Figure S2: Determination of the optimal number of RGs in control PCLS by geNorm analysis; Figure S3: Determination of the optimal number of RGs in treated PCLS by geNorm analysis; Figure S4: Determination of the optimal number of RGs in control PCLS by geNorm analysis; Figure S5: Determination
of the optimal number of RGs in treated PCLS by geNorm analysis; Figure S6: Determination of the optimal number of RGs in the control PCLS by geNorm analysis; Figure S7: Determination of the optimal number of RGs in treated PCLS by geNorm analysis; Table S1: MIQE checklist; Table S2: Summary of RIN values; Table S3: Summary of liver biochemical parameters; Table S4: RefFinder results for mRNA of human liver tissues (n=35); Table S5: Effects of different normalization approaches on the expression of CYP1A2 and CYP3A4. Results are presented as 2-ΔΔCq values relative to sample L1.; Table S6: RefFinder results for the mRNA of the control PCLS incubated for 24 h with DMSO; Table S7: RefFinder results for the mRNA of the treated PCLS incubated for 24 h with RIF/BNF/DMSO; Table S8: RefFinder results for the mRNA of the control PCLS incubated for 24 h with DMSO (time dependence); Table S9: RefFinder results for the mRNA of the treated PCLS incubated for 24 h with RIF/BNF/DMSO (time dependence); Table S10: RefFinder results for the mRNA of the control PCLS incubated for 24 h with DMSO (time dependence); Table S11: RefFinder results for the mRNA of the treated PCLS incubated for 24 h with RIF/BNF/DMSO (time dependence).

Author Contributions: Conceptualization, P.M. and I.B.; Methodology, T.Z., M.A., Z.Š., and P.M.; Validation, T.Z., M.A., Z.Š., and P.M.; Formal analysis, T.Z. and P.M.; Investigation, T.Z., M.A., and Z.Š.; Resources, Z.Š., I.B., and L.S.; Data curation, T.Z., Z.Š., and P.M.; Writing, original draft preparation, T.Z.; Writing – review & editing, I.B., P.M., and L.S.; Visualization, T.Z.; Supervision, I.B. and L.S.; Project administration, I.B.; Funding acquisition, I.B. and L.S. All authors read and approved the final manuscript.

Funding: This research was funded by the Czech Science Foundation (Grant No. 18-09946S) and by Charles University (Research Project SVV 260 416). Lenka Skalová was partly supported by the project EFSA-CDN [CZ.02.1.01/0.0/0.0/16_019/000841], co-funded by European Regional Development Fund (ERDF).

Acknowledgments: The authors thank Daniel Paul Sampey for the English revision.

Conflicts of Interest: The authors declare no conflicts of interest.

References
1. Fisher, R.L.; Vickers, A.E.M. Preparation and culture of precision-cut organ slices from human and animal. Xenobiotica. 2013, 43, 8–14. [CrossRef] [PubMed]
2. Olinga, P.; Hof, I.H.; Merema, M.T.; Smit, M.; de Jager, M.H.; Swart, P.J.; Sloff, M.J.H.; Meijer, D.K.F.; Groothuis, G.M.M. The applicability of rat and human liver slices to the study of mechanisms of hepatic drug uptake. J. Pharmacol. Toxicol. Methods. 2001, 45, 35–63. [CrossRef]
3. Vickers, A.E.M.; Ulyanov, A.V.; Fisher, R.L. Liver Effects of Clinical Drugs Differentiated in Human Liver Slices. Int. J. Mol. Sci. 2017, 18, 33. [CrossRef] [PubMed]
4. Zarybnicky, T.; Matouskova, P.; Lancosova, B.; Subrt, Z.; Skalova, L.; Bousova, I. Inter-Individual Variability in Acute Toxicity of R-Pulegone and R-Menthofuran in Human Liver Slices and Their Influence on miRNA Expression Changes in Comparison to Acetaminophen. Int. J. Mol. Sci. 2018, 19, 17. [CrossRef] [PubMed]
5. Rius, B.; Titos, E.; Moran-Salvador, E.; Lopez-Vicario, C.; Garcia-Alonso, V.; Gonzalez-Periz, A.; Arroyo, V.; Claria, J. Resolvin D1 primes the resolution process initiated by calorie restriction in obesity-induced steatohepatitis. PLoS Biol. 2014, 28, 836–848. [CrossRef] [PubMed]
6. Rius, B.; Duran-Guell, M.; Flores-Costa, R.; Lopez-Vicario, C.; Lopategi, A.; Alcaraz-Quiles, J.; Casulleras, M.; Lozano, J.J.; Titos, E.; Claria, J. The specialized proresolving lipid mediator maresin 1 protects hepatocytes from lipotoxic and hypoxia-induced endoplasmic reticulum stress. PLoS Biol. 2017, 31, 5384–5398. [CrossRef] [PubMed]
7. Wu, X.; Roberto, J.B.; Knupp, A.; Kenerson, H.L.; Truong, C.D.; Yuen, S.Y.; Brempeilis, K.J.; Tuefferd, M.; Chen, A.; Horton, H.; et al. Precision-cut human liver slice cultures as an immunological platform. J. Immunol. Methods. 2018, 455, 71–79. [CrossRef]
8. Elferink, M.G.L.; Olinga, P.; Draaisma, A.L.; Merema, M.T.; Bauerschmidt, S.; Polman, J.; Schoonen, W.G.; Groothuis, G.M.M. Microarray analysis in rat liver slices correctly predicts in vivo hepatotoxicity. Toxicol. Appl. Pharmacol. 2008, 229, 300–309. [CrossRef]
9. Melgert, B.N.; Olinga, P.; Van der Laan, J.M.S.; Weert, B.; Cho, J.; Schuppan, D.; Groothuis, G.M.M.; Meijer, D.K.F.; Poelstra, K. Targeting dexamethasone to Kupffer cells: Effects on liver inflammation and fibrosis in rats. Hepatology 2001, 34, 719–728. [CrossRef]
10. van de Bovenkamp, M.; Groothuis, G.M.M.; Draaisma, A.L.; Merema, M.T.; Bezuijen, J.I.; van Gils, M.J.; Meijer, D.K.F.; Friedman, S.L.; Olinga, P. Precision-cut liver slices as a new model to study toxicity-induced hepatic stellate cell activation in a physiologic milieu. Toxicol. Sci. 2005, 85, 632–638. [CrossRef]
11. Vickers, A.E.M.; Saulnier, M.; Cruz, E.; Merema, M.T.; Rose, K.; Bentley, P.; Olinga, P. Organ slice viability extended for pathway characterization: An in vitro model to investigate fibrosis. *Toxicol. Sci.* 2004, 82, 534–544. [CrossRef]

12. Starokozhko, V.; Vatakuti, S.; Schievink, B.; Merema, M.T.; Asplund, A.; Synnergren, J.; Aspegren, A.; Groothuis, G.M.M. Maintenance of drug metabolism and transport functions in human precision-cut liver slices during prolonged incubation for 5 days. *Arch. Toxicol.* 2017, 91, 2079–2092. [CrossRef]

13. Vandesompele, J.; De Preter, K.; Pattyn, F.; Poppe, B.; Van Roy, N.; De Paepe, A.; Speleman, F. Accurate normalization of real-time quantitative RT-PCR data by geometric averaging of multiple internal control genes. *Genome Biol.* 2002, 3, 12. [CrossRef]

14. Bustin, S.A. Quantification of mRNA using real-time reverse transcription PCR (RT-PCR): Trends and problems. *J. Mol. Endocrinol.* 2002, 29, 17. [CrossRef]

15. Vickers, A.E.M.; Fisher, R.L. Human Liver Slices to Investigate Injury and Repair. *Appl. In Vitro Toxicol.* 2018, 4, 280–287. [CrossRef]

16. Wan, H.; Zhao, Z.; Qian, C.; Sui, Y.; Malik, A.A.; Chen, J. Selection of appropriate reference genes for gene expression studies by quantitative real-time polymerase chain reaction in cucumber. *Anal. Biochem.* 2010, 399, 257–261. [CrossRef]

17. Bustin, S.A.; Benes, V.; Garson, J.A.; Hellemans, J.; Huggett, J.; Kubista, M.; Mueller, R.; Nolan, T.; Pfaff, M.W.; Shipley, G.L.; et al. The MIQE Guidelines: Minimum Information for Publication of Quantitative Real-Time PCR Experiments. *Clin. Chem.* 2009, 55, 611–622. [CrossRef]

18. Bustin, S.; Nolan, T. Talking the talk, but not walking the walk: RT-qPCR as a paradigm for the lack of reproducibility in molecular research. *Eur. J. Clin. Invest.* 2017, 47, 756–774. [CrossRef]

19. Gerets, H.H.J.; Tilmant, K.; Gerin, B.; Chanteux, H.; Depelchin, B.O.; Dhalluin, S.; Atienzar, F.A. Characterization of primary human hepatocytes, HepG2 cells, and HepaRG cells at the mRNA level and CYP activity in response to inducers and their predictivity for the detection of human hepatotoxins. *Cell Biol. Toxicol.* 2012, 28, 69–87. [CrossRef]

20. Hadi, M.; Westra, I.M.; Starokozhko, V.; Dragovic, S.; Merema, M.T.; Groothuis, G.M.M. Human Precision-Cut Liver Slices as an ex Vivo Model to Study Idiosyncratic Drug-Induced Liver Injury. *Chem. Res. Toxicol.* 2013, 26, 710–720. [CrossRef]

21. Untergasser, A.; Cutcutache, I.; Koressaar, T.; Ye, J.; Faircloth, B.C.; Remm, M.; Rozen, S.G. Primer3—new capabilities and interfaces. *Nucleic Acids Res.* 2012, 40, e115. [CrossRef]

22. Zuker, M. Mfold web server for nucleic acid folding and hybridization prediction. *Nucleic Acids Res.* 2003, 31, 3406–3415. [CrossRef]

23. Lardizabal, M.N.; Nocito, A.L.; Daniele, S.M.; Ornella, L.A.; Palatnik, J.F.; Veggi, L.M. Reference Genes for Real-Time PCR Quantification of MicroRNAs and Messenger RNAs in Rat Models of Hepatotoxicity. *PLoS One.* 2012, 7, e36323. [CrossRef]

24. Chen, C.; Ridzon, D.A.; Broomer, A.J.; Zhou, Z.; Lee, D.H.; Nguyen, J.T.; Barbisin, M.; Xu, N.L.; Mahuvakar, V.R.; Andersen, M.R.; et al. Real-time quantification of microRNAs by stem–loop RT–PCR. *Nucleic Acids Res.* 2005, 33, e179. [CrossRef]

25. Kibbe, W.A. OligoCalc: An online oligonucleotide properties calculator. *Nucleic Acids Res.* 2007, 35, W43–W46. [CrossRef]

26. Pfaffl, M.W.; Tichopad, A.; Prgovc, C.; Neuvians, T.P. Determination of stable housekeeping genes, differentially regulated target genes and sample integrity: BestKeeper - Excel-based tool using pair-wise correlations. *Biotechnol. Lett.* 2004, 26, 509–515. [CrossRef]

27. Andersen, C.L.; Jensen, J.L.; Orntoft, T.F. Normalization of real-time quantitative reverse transcription-PCR data: A model-based variance estimation approach to identify genes suited for normalization, applied to bladder and colon cancer data sets. *Cancer Res.* 2004, 64, 5245–5250. [CrossRef]

28. Silver, N.; Best, S.; Jiang, J.; Thein, S.L. Selection of housekeeping genes for gene expression studies in human reticulocytes using real-time PCR. *BMC Mol. Biol.* 2006, 7, 9. [CrossRef]

29. Xie, F.L.; Xiao, P.; Chen, D.L.; Xu, L.; Zhang, B.H. miRDeepFinder: A miRNA analysis tool for deep sequencing of plant small RNAs. *Plant Mol.Biol.* 2012, 80, 75–84. [CrossRef]

30. Livak, K.J.; Schmittgen, T.D. Analysis of Relative Gene Expression Data Using Real-Time Quantitative PCR and the 2−∆∆CT Method. *Methods.* 2001, 25, 402–408. [CrossRef]
31. Pavo, N.; Raderer, M.; Goliasch, G.; Wurm, R.; Strunk, G.; Cho, A.; Novak, J.F.; Gisslinger, H.; Steger, G.G.; Hejna, M.; et al. Subclinical involvement of the liver is associated with prognosis in treatment naïve cancer patients. *Oncotarget*. 2017, 8, 81250–81260. [CrossRef] [PubMed]

32. Nygard, A.B.; Jorgensen, C.B.; Cirera, S.; Fredholm, M. Selection of reference genes for gene expression studies in pig tissues using SYBR green qPCR. *BMC Mol. Biol.* 2007, 8, 67. [CrossRef] [PubMed]

33. Ahn, K.; Huh, J.W.; Park, S.J.; Kim, D.S.; Ha, H.S.; Kim, Y.J.; Lee, J.R.; Chang, K.T.; Kim, H.S. Selection of internal reference genes for SYBR green qRT-PCR studies of rhesus monkey (Macaca mulatta) tissues. *BMC Mol. Biol.* 2008, 9, 78. [CrossRef] [PubMed]

34. Granitzny, A.; Knebel, J.; Schaudien, D.; Braun, A.; Steinberg, P.; Dasenbrock, C.; Hansen, T. Maintenance of high quality rat precision cut liver slices during culture to study hepatotoxic responses: Acetaminophen as a model compound. *Toxicol. In Vitro* 2017, 42, 200–213. [CrossRef] [PubMed]

35. Lee, S.M.L.; Schelcher, C.; Gashi, S.; Schreiber, S.; Thasler, R.M.K.; Jauch, K.-W.; Thasler, W.E. RNA Stability in Human Liver: Comparison of Different Processing Times, Temperatures and Methods. *Mol. Biotechnol.* 2013, 53, 1–8. [CrossRef] [PubMed]

36. Vickers, A.E.; Fisher, R.L.; Sinclair, J.R. Glutathione Modulation and Oxidative Stress in Human Liver Slices. *Curr. Drug Discov. Technol.* 2010, 7, 154–169. [CrossRef] [PubMed]

37. Bartel, D.P. MicroRNAs: Target Recognition and Regulatory Functions. *Cell* 2009, 136, 215–233. [CrossRef] [PubMed]

38. Matoušková, P. microRNAs and Reference Gene Methodology. In *Handbook of Nutrition, Diet, and Epigenetics*, 1st ed.; Patel, V., Preedy, V., Eds.; Springer International Publishing: Cham, Switzerland, 2017; pp. 1–17.

39. Wang, J.; Lu, M.; Qiu, C.; Cui, Q. TransmiR: A transcription factor–microRNA regulation database. *Nucleic Acids Res.* 2010, 38, D119–D122. [CrossRef] [PubMed]

40. Schwarzenbach, H.; da Silva, A.M.; Calin, G.; Pantel, K. Data Normalization Strategies for MicroRNA Quantification. *Clin. Chem.* 2015, 61, 1333–1342. [CrossRef] [PubMed]

41. Matoušková, P.; Bártíková, H.; Bousová, I.; Hanušová, B.; Skálová, L. Reference Genes for Real-Time PCR Quantification of Messenger RNAs and MicroRNAs in Mouse Model of Obesity. *PLoS ONE* 2014, 9, e86033. [CrossRef]

42. Lamba, V.; Ghodke-Puranik, Y.; Guan, W.; Lamba, J.K. Identification of suitable reference genes for hepatic microRNA quantitation. *BMC Res. Notes* 2014, 7, 129. [CrossRef]

43. Zhang, J.; Le, T.D.; Liu, L.; Liu, B.; He, J.; Goodall, G.J.; Li, J. Identifying direct miRNA–mRNA causal regulatory relationships in heterogeneous data. *J. Biomed. Inform.* 2014, 52, 438–447. [CrossRef]

44. Tang, X.; Chen, S. Epigenetic Regulation of Cytochrome P450 Enzymes and Clinical Implication. *Curr. Drug Metab.* 2015, 16, 86–96. [CrossRef]

45. Ramamoorthy, A.; Liu, Y.; Philips, S.; Desta, Z.; Lin, H.; Goswami, C.; Gaedigk, A.; Li, L.; Flockhart, D.A.; Skaar, T.C. Regulation of MicroRNA Expression by Rifampin in Human Hepatocytes. *Drug Metab. Dispos.* 2013, 41, 1763–1768. [CrossRef]

46. Takahashi, K.; Tatsumi, N.; Fukami, T.; Yokoi, T.; Nakajima, M. Integrated Analysis of Rifampicin-induced MicroRNA and Gene Expression Changes in Human Hepatocytes. *Drug Metab. Pharmacokinet.* 2014, 29, 333–340. [CrossRef]

47. Dheda, K.; Huggett, J.F.; Chang, J.S.; Kim, L.U.; Bustin, S.A.; Johnson, M.A.; Rook, G.A.W.; Zumla, A. The implications of using an inappropriate reference gene for real-time reverse transcription PCR data normalization. *Anal. Biochem.* 2005, 344, 141–143. [CrossRef]