The vital role of platelets in hemostasis was recognized >130 years ago. Nowadays, it is established that platelets are involved not only in many more physiological but also in pathophysiological processes: they act as safeguards of vascular integrity and contribute to inflammation, tumor metastasis, liver regeneration and the genesis, and progression of cardiovascular diseases. Both, the absence of a nucleus (and consequently limited levels of protein synthesis) and the regulation of platelet activity at the level of post-translational modifications (PTM), render proteomics an invaluable tool for characterizing the fundamental processes that affect platelet homeostasis and thus determine the roles of platelets in health and disease. In this article we provide a critical overview on the achievements, the current possibilities, and the future perspectives of platelet proteomics to study patients experiencing cardiovascular, inflammatory, and bleeding disorders. (Circ Res. 2014;114:1204-1219.)

Key Words: bleeding ■ blood platelets ■ cardiovascular diseases ■ hemorrhage ■ proteome ■ proteomics

What Can Proteomics Tell Us About Platelets?

Julia M. Burkhart, Stepan Gambaryan, Stephen P. Watson, Kerstin Jurk, Ulrich Walter, Albert Sickmann, Johan W.M. Heemskerk, René P. Zahedi

Abstract: More than 130 years ago, it was recognized that platelets are key mediators of hemostasis. Nowadays, it is established that platelets participate in additional physiological processes and contribute to the genesis and progression of cardiovascular diseases. Recent data indicate that the platelet proteome, defined as the complete set of expressed proteins, comprises >5000 proteins and is highly similar between different healthy individuals. Owing to their anucleate nature, platelets have limited protein synthesis. By implication, in patients experiencing platelet disorders, platelet (dys)function is almost completely attributable to alterations in protein expression and dynamic differences in post-translational modifications. Modern platelet proteomics approaches can reveal (1) quantitative changes in the abundance of thousands of proteins, (2) post-translational modifications, (3) protein–protein interactions, and (4) protein localization, while requiring only small blood donations in the range of a few milliliters. Consequently, platelet proteomics will represent an invaluable tool for characterizing the fundamental processes that affect platelet homeostasis and thus determine the roles of platelets in health and disease. In this article we provide a critical overview on the achievements, the current possibilities, and the future perspectives of platelet proteomics to study patients experiencing cardiovascular, inflammatory, and bleeding disorders. (Circ Res. 2014;114:1204-1219.)

Key Words: bleeding ■ blood platelets ■ cardiovascular diseases ■ hemorrhage ■ proteome ■ proteomics

The online-only Data Supplement is available with this article at http://circres.ahajournals.org/lookup/suppl/doi:10.1161/CIRCRESAHA.114.301598/-/DC1.

© 2014 American Heart Association, Inc.

Circulation Research is available at http://circres.ahajournals.org

DOI: 10.1161/CIRCRESAHA.114.301598
characterizing the fundamental processes that affect platelet homeostasis and thus determine the roles of platelets in health and disease. Moreover, platelets can easily be isolated from the human body as a relatively pure and homogeneous population of primary cells. The great variety in functions beyond hemostasis renders them—and hence platelet proteomics—interesting for different application fields, where the use of genomic approaches might be limited. In this article we provide a critical overview about the achievements, the current possibilities, and the future perspectives of platelet proteomics.

Current State of Platelet Proteomics

The proteome can be described as the complete set of proteins expressed within a defined sample under distinct conditions. As basically all biological processes are regulated by proteins, the proteome—in contrast to the genome—is highly dynamic and can change its composition both qualitatively and quantitatively over time. Recent studies demonstrated that using modern proteomics technology, ~10,000 proteins can be detected in human cell lines, covering a dynamic range of 7 or more orders of magnitude. More than 300 known PTM patterns considerably narrow its application range: (1) Large, small, hydrophobic, and basic proteins cannot be resolved well; (2) the sensitivity and linear/dynamic range depend on the applied staining procedure; (3) before MS, each protein spot of interest has to be excised, washed, and subjected to in-gel proteolytic digestion, thus reducing the detection limit at the expense of cost/time per protein identification; (4) 2-DE quantification is based on spot intensities after staining, but in complex samples, spots usually contain more than a single protein, rendering quantification prone to errors; and (5) the availability of only a few identifications can impede a thorough validation to estimate the share of false-positive protein identifications. However, 2-DE has several inherent limitations that considerably narrow its application range: (1) Large, small, hydrophobic, and basic proteins cannot be resolved well; (2) the sensitivity and linear/dynamic range depend on the applied staining procedure; (3) before MS, each protein spot of interest has to be excised, washed, and subjected to in-gel proteolytic digestion, thus reducing the detection limit at the expense of cost/time per protein identification; (4) 2-DE quantification is based on spot intensities after staining, but in complex samples, spots usually contain more than a single protein, rendering quantification prone to errors; and (5) the availability of only a few identifications can impede a thorough validation to estimate the share of false-positive protein identifications. However, 2-DE can resolve protein isoforms, and in contrast to liquid chromatography (LC)-MS, the proteomic characterization of larger cohorts using 2-DE has been established for many years.

Most of the aforementioned limitations of 2-DE are overcome with the advent of advanced gel-free proteomics techniques. This in particular concerns 3 developments: (1) availability of improved mass spectrometers providing higher sensitivity, specificity (mass accuracy, resolution), and speed/throughput (as well as novel scanning techniques); (2) better high performance LC systems and materials providing an enhanced separation of highly complex samples; and (3) novel bioinformatics strategies enabling highly complex data interpretation at well-defined quality measures. Yet, these technical developments require robust protocols and extensive quality control to translate big data into reasonable results. Using state-of-the-art LC-MS bottom-up setups (Text Box 1), it is possible to identify and quantify thousands of proteins within hours at a given quality level, usually referred to as false discovery rate (FDR, Text Box 1), representing an estimated share of unwanted but still inevitable random identifications. One limitation of these bottom-up approaches is that information about protein isoforms can be lost because identification and quantification are conducted on the peptide level.

Owing to the continuous propagation of MS-based proteomics throughout life sciences, there has been a tremendous development of approaches to analyze proteins, their abundances, their interaction partners, their mature N and C termini, as well as their PTM patterns. Although several of these MS-based techniques have been used to deepen our understanding of human platelets (Figure 1), we are only beginning to exploit the range of possibilities, which is far beyond the mere identification of the most abundant proteins.

Depending on the study design, different types of information can be obtained from platelet proteome analyses (Figure 1). Proteomics can provide detailed insights into the (quantitative) protein composition of platelets, of their subcompartments (eg, the plasma membrane) or platelet-derived microparticles, examined at resting or activated conditions. In addition, proteomics can provide quantitative information about changes in protein and PTM levels, for instance on perturbation of the resting state (eg, agonist treatment) or when comparing platelets derived from different donors (genomic background or healthy versus disease). Accordingly, platelet proteome studies cannot only contribute to revealing functional mechanisms but also help to identify target proteins that associate with the genesis or progression of a pathological condition. Such studies thus promise to provide potential biomarkers for prevalent diseases. Yet, examples of proteomics-derived biomarkers, which have made it into application, are scarce. However, the steady and fast progress of proteomic technologies in the past few years has paved a way toward application in clinical research and holds a great potential for the future. One big challenge for platelet (biomarker) analysis still is preanalytical interference for instance because of impurity or preactivation as the levels of known platelet biomarkers are known to be influenced by platelet isolation protocols. To circumvent artificial results, there is an urgent need for robust and standardized protocols of platelet preparation and activation, as well as protein isolation.

Quantitative Human Platelet Proteome

Owing to their lineage as anucleate fragments of megakaryocytes and their unique role, it is evident that platelets differ considerably from nucleate cells and cultured cell lines. Platelets can be isolated as a relatively homogenous population of primary (human) cells, and hence, corresponding data
**Text Box 1. Liquid Chromatography-Mass Spectrometry Analysis**

Although also intact proteins and complexes can be analyzed by liquid chromatography-mass spectrometry (MS), the method of choice in MS-based proteomics is still so-called bottom-up, in which proteins are enzymatically cleaved into peptides (commonly with trypsin) before analysis, leading to multiplication of sample complexity from 1000s of proteins to 100 000s of peptides. Advantages are that, for peptides, (1) separation by chromatography, (2) ionization, and (3) fragmentation are more efficient, (4) their masses are less heterogeneously distributed, and (5) their identification through database searches is more straightforward. Complex peptide mixtures can be either directly analyzed by liquid chromatography-MS using long gradients or fractionated by hyphenation of different separation techniques beforehand to reduce sample complexity and increase depth and coverage of the proteome analysis. Peptide sequences are usually identified by comparing the molecular masses of the peptide (parent) ion and corresponding fragment ions, as determined in MS and MS/MS scans, with the predicted peptide/fragment masses obtained from in silico digestion of a protein database. Nowadays, results are reported at defined false discovery rates, representing the share of random hits within large data sets. Therefore, for a selected protein database (eg, uniprot human or mouse), all sequences are usually reversed to generate a concatenated target/decoy (forward/reverse) database. On searching MS data against such a database, the number of decoy hits represents the number of random hits in the target database, such that the share of false-positive hits can be assessed on the peptide spectrum match, the peptide, and the protein level. However, the validity of the underlying statistics depends on the availability of sufficient data points (peptide spectrum match, peptide, and protein identifications). Notably, the presence of genetic polymorphisms still poses a challenge for global proteomic workflows because they can alter peptide and fragment ion masses. Nevertheless, they can contribute to platelet dysfunction and, therefore, will represent an important field for future platelet proteomics studies.

For the past decade, MS has been the method of choice to identify and quantify post-translational modifications (PTM) in proteins and peptides, given that those induce a detectable mass shift and are either stable during sample preparation and MS detection, or, before analysis, can be chemically/enzymatically modified for this purpose. Notably, a direct detection of the intact PTM is preferable to indirect methods, which rely on induction of artificial mass shifts providing a more confident way of identification. In principle, MS can identify the presence of a PTM but also pinpoint its distinct localization within the peptide sequence. As common search algorithms for peptide identification are prone to errors if modifiable amino acids are in close proximity, the use of specific algorithms to assess the quality and confidence of PTM localizations is mandatory. Because the implementation of such algorithms is relatively new, it should be kept in mind that public PTM databases might contain a certain level of wrong annotations, especially in case of older data sets.

Despite the relevance of knowledge of PTM, their sheer versatility and usually low stoichiometry render the usage of specific enrichment strategies for removal of the bulk of nonmodified peptides inevitable. However, especially when PTM-directed antibodies are used for the analysis of ubiquitination or acetylation, sensitivity and consequently large sample amounts can be an issue, even requiring several milligrams per sample for a comprehensive analysis.

May yield more relevant insights compared with cultured and endogenous cells with major differences in differentiation stages. Detailed knowledge of the protein composition of platelets and the differences compared with nucleated cells is, therefore, essential to improve our understanding of the delicate equilibrium between platelet inhibition and activation. Platelets are isolated from either freshly donated blood or leukocyte-depleted stored platelet concentrates. Whereas the latter is more convenient, fresh blood donations are preferable to ensure that the sample is as close to the in vivo situation as possible. For a proper analysis, contamination of erythrocytes, leukocytes, circulating stem cells, and plasma have to be minimized and preferentially monitored, as these clearly bias the final results, especially given the 7 orders of magnitude range of protein expression levels in human cells. This leads to a trade-off between effort and feasibility on the one side and reproducibility and quality on the other side.

Taking this into account, we recently published a comprehensive study to obtain a quantitative map of the human platelet proteome landscape. The rationale was to establish an optimized and reproducible protocol for isolation of washed platelets with a clear focus on purity and to analyze the proteome of the isolated platelets thoroughly, taking into account additional purification steps and sample losses. We identified 4000 individual proteins (with an FDR on the protein level of <1%) and, using spectral counting, we compared our MS data with copy numbers of 24 proteins found in the literature. Based on the high degree of correlation (coefficient of determination \(R^2 = 0.9\)), we provided copy number estimates for \(=3600\) proteins thus vastly expanding the knowledge about the composition of human platelets. Based on the coverage of known pathways on the one hand and the agreement of calculated (quantitative MS, 1.5 mg) and measured (laboratory values, 1.8±0.2 mg) total protein mass per 10^9 platelets on the other hand, we estimated that these data represent >80% of the complete platelet proteome. From our quantitative MS data, we could furthermore deduce that, using our reproducible isolation procedure, >80% of the platelet proteome showed only minor inter- and intradonor variation and were stable (1) between 4 healthy donors and (2) between blood donations of 1 donor over time (≈3 weeks). This was in good agreement with 2-DE derived data from Winkler et al, who found a coefficient of variation of 18% based on 500 spots reproducibly found in platelets from healthy donors. In our study, the calculated amount of the 2 most prominent plasma proteins, serum albumin and transthyretin, allowed measurement of the degree of plasma contamination, which is inevitable because of the sponge-like structure of the platelet open canalicular system. The well-purified, washed platelet samples contained only 1.5% by volume of plasma, which was lower than in previous platelet proteome studies.

Notably, for differential studies focusing on changes in the platelet proteome, for example, in disease or on drug intake, and for which the accessible blood volume might be limited, the reproducibility of the platelet preparation is as important as its purity. If the level of nonplatelet proteins remains constant but limited, contamination by other blood components can be tolerated. For high reproducibility and confidence of
the generated data, it is important to monitor the entire workflow including (1) patient selection, (2) blood drawing, (3) platelet isolation and purity, (4) lysis and proteolytic digestion, along with analytic steps such as (5) chemical labeling, (6) prefractionation of the peptide mixture, (7) LC-MS analysis, and (8) MS data interpretation.

At present, costs and effort to conduct LC-MS–based proteome-wide analyses from individual patients are relatively high. These require state-of-the-art equipment and either the usage of costly stable isotope labeling techniques which allow multiplexing of ≤10 samples to save time or the successive analysis of single samples by label-free quantification strategies to characterize larger cohorts (Text Box 1). In the past few years, 2-DE was preferably used for disease-driven platelet proteomics, but as mentioned above, this technique has limitations when compared with gel-free LC-MS–based proteomics (Table I in the online Data Supplement).

Because of the rapid progress in the development of MS instrumentation, costs and time per identified (and quantified) protein will reduce further, but it is questionable whether this will equally apply for stable isotope labeling reagents. Alternatively, so-called targeted LC-MS assays, such as multiple reaction monitoring and high-resolution targeted-MS/MS, allow fast, accurate, and sensitive quantification of predefined sets of proteins in complete cell lysates (Figure 2; Text Box 2). These techniques have already been used to monitor protein signatures in a variety of studies and might be suitable to monitor defined proteomic changes in platelets from larger patient cohorts.

In this context, recently introduced SWATH-MS clearly has good prospects. Here, based on a priori information from spectral libraries, samples are analyzed in a data-independent manner, such that all ions present are fragmented in consecutive 25-Da windows. This allows for quantification of many more analytes than with conventional targeted methods with similar accuracy, however less sensitivity. One drawback might be the dependence on spectral libraries for precise data interpretation, which is not required in other data-independent strategies.

**Platelet Proteome and Transcriptome**

In support of the observations that essential components of the RNA translational machinery are present in platelets, microarray-based studies discovered that ≤32% of all...
human genes are expressed in platelets at the messenger RNA level. Several studies focused on analysis of the platelet transcriptome and some of these compared protein and transcript levels, deducing a certain degree of correlation. A comprehensive comparison between our quantitative proteome data and published transcriptome data revealed only weak correlation. The messenger RNA found in platelets originates from megakaryocytes and is likely to be affected by guest on July 9, 2017 http://circres.ahajournals.org/ Downloaded from

**Text Box 2. On the Use of Stable Isotopes**

Incorporation of heavy stable isotopes (13C, 15N, 18O, and 2H) is a widely used strategy for quantification in mass spectrometry (MS)-based proteomics. For the hypothetical peptide sequence PLATELET PROTEOMICS, the light (endogenous) and the heavy stable isotope-coded versions (eg, containing 13C6-labeled Arg) have the same chemical properties but differ in physical properties. They thus behave equally during sample preparation, liquid chromatography separation, and MS ionization. Yet, based on the 6-Da mass shift derived from the presence of 6 additional neutrons in the heavy version, both forms are distinguished by MS, and their signal intensities can be used to determine their relative ratio. Incorporation of stable isotopes is achieved by chemical and metabolic labeling, but use of the latter method is limited to cell culture and some model organisms, among those the SILAC (Stable isotope labeling by amino acids in cell culture) mouse. In case of isobaric chemical labeling techniques such as isobaric tag for relative and absolute quantitation (iTRAQ) and tandem mass tag (TMT), all differently labeled forms of a peptide have the same mass, as represented by a single signal on the MS1 level (Figure 2). Once the peptide is isolated and fragmented, reporter ions corresponding to relative amounts of the peptide within the labeled samples are released for each label and detected in a subsequent MS/MS scan. iTRAQ and TMT currently allow multiplexing of ≤8 and 10 samples, respectively.

Reductive dimethyl labeling is a comparatively cheap alternative, however, at the expense of several disadvantages when compared with the more costly iTRAQ and TMT technologies. (1) Currently multiplexing is limited to only 3 samples and, as for all nonisobaric labels, 2-plexing will double and 3-plexing will triple the complexity of the sample. (2) Incorporation of deuterium causes retention time shifts between light/medium/heavy forms of the peptide during reversed phase chromatography and thus can complicate quantification by liquid chromatography-MS. Hebert et al recently introduced a novel strategy for neutron-encoded labeling, which will increase the multiplexing capabilities of the aforementioned labeling strategies, but requires mass spectrometers with a resolution power of ≈500,000 and above, currently supported by only few instruments in conjunction with reasonable sensitivity and scanning speed.

**Figure 2. Typical liquid chromatography (LC)-mass spectrometry (MS) based proteomic workflows.** A, Sample preparation. Proteins are extracted and enzymatically cleaved into peptides (1–4). Modified peptides need to be specifically enriched in case of post-translational modification (PTM) analysis. B, LC-MS analysis. For discovery, qualitative or label-free quantitative (*) analysis of individual samples (5) or stable isotope labeling (SIL) of peptides (6) with subsequent multiplexing and analysis of pooled samples can be applied. Both strategies can be combined with fractionation to reduce sample complexity and to increase proteome coverage of the LC-MS analysis (7) and following data interpretation. Stable isotope labeling can be performed with nonisobaric and isobaric labels (Text Box 2). Whereas nonisobaric labeling requires quantification based on parent ions (MS1 level) and leads to duplication/triplication of sample complexity. isobaric labels are quantified on the level of fragment ions (MS/MS) and thus allow multiplexing of ≤10 samples without considerable increase in sample complexity. For validation and screening, specific approaches such as targeted MS/MS can be applied (8), allowing for a sensitive detection and quantification of specific peptides/proteins in complete cell lysates. Demonstrated is the detection of a peptide from sortilin (~700 copies per platelet) with high signal-to-noise, obtained by a 2-hour LC-MS analysis of 1 µg of complete platelet lysate, which consists of >5000 different proteins, covering ~5 orders of magnitude. TIC indicates total ion chromatogram; and UV, ultraviolet.
by aging and platelet activation. Moreover, platelets can exchange RNA with other cells.

A recent study detected 532 mature micro-RNA species in resting platelets. Micro-RNAs are known to be involved in diverse processes such as myeloid cell differentiation, megakaryocytogenesis, and thrombopoiesis, and the authors hypothesize that micro-RNAs might serve as biomarkers for platelet-related disorders, although this remains to be proven.

Platelet Subproteomes

A variety of studies has focused on the characterization of platelet subproteomes. In contrast to the global analyses, subproteomes can provide insights into the subcellular localization or PTM patterns of proteins. Subproteome analyses can reveal (patho)physiologically important processes and, therefore, hold the potential for unraveling previously unknown mechanisms. Figure 3 gives a summary of published platelet (sub)proteome studies.

The platelet plasma membrane represents the interface for interaction with other blood components and transmits responses to external stimuli. Many drug targets are membrane proteins, and thus there is considerable interest in the plasma membrane subproteome. Dedicated studies used gel-based or gel-free approaches, with the latter yielding ≈650 membrane proteins, identifying even low abundant 7 transmembrane domain receptors, such as the P2Y1 receptor for ADP (≈150 copies per platelet). These studies clearly extended our knowledge of plasma membrane proteins and receptors. Despite the increasing power of proteomic workflows, membrane proteins can easily escape detection and thus may be under-represented. Especially 2-DE is not well-suited for analysis of hydrophobic membrane proteins, which often are quantitatively lost. Even for gel-free strategies, the hydrophobic nature of membrane proteins renders sample preparation challenging. For instance, the absence of trypsin cleavage sites within transmembrane domains may require the use of alternative enzymes/strategies. In addition, frequently present glycosylation can hamper proteolytic digestion and also impede identification by MS. Despite these problems, ≈75% of the known platelet membrane receptors have been identified in the 2 to date most comprehensive LC-MS–based proteomic studies (Table 1).

Platelet-derived microparticles are common components of the blood, which are increasingly released on aging and also during activation of platelets, increasing in number under pathological conditions. Several hundreds of proteins could be identified in platelet microparticles after gel electrophoresis and LC-MS analysis of 26 excised bands. A preliminary analysis showed that platelet microparticles can be separated in size classes with different protein composition; however, the applied data interpretation does not meet current standards in the field. Recently, microparticles induced by thrombin or shear were compared using 2-DE, resulting in 26 differential proteins.

Activation of platelets triggers α-granule secretion, leading to the redistribution of granule membrane proteins to the plasma membrane and the release of soluble proteins into the blood. Several studies analyzed the proteome of α-granules isolated from platelets. Wijten et al. used a quantitative gel-free proteomics approach to compare the proteome of collagen- and thrombin-stimulated platelets with that of unstimulated platelets to identify proteins that are downregulated because of secretion. The authors describe a high correlation between their platelet proteome data and our previously published quantitative platelet proteome and could identify 124 released proteins. However, already in 2004, Coppinger et al. identified >300 proteins after MS analysis of a platelet releasate depleted from microparticles and intact cells. One reason for this discrepancy might be contamination derived from unanticipated platelet lysis as Wijten et al. analyzed changes in the proteome of intact platelets rather than in the releasate itself. Moreover, Coppinger et al. used an elaborate data interpretation workflow, but no methods for estimating FDRs were available at that time. Nowadays, proteomics data are usually provided with a defined FDR (routinely 1%), which results in an improved validity, usually at the expense of reduced protein numbers.

Activity-based proteomics technologies use active site–directed chemical probes, which can specifically and irreversibly modify enzyme active sites (eg, for subsequent enrichment and direct analysis of enzyme activity, respectively). Activity-based protein profiling approaches have been used to analyze ATP-binding proteins in platelets and to enrich cyclic nucleotide–binding proteins (discussed in a recent review by Holly et al.). In one study, proteins from resting and collagen receptor–stimulated platelets were incubated with cAMP/cGMP-coupled beads, and the pull-down eluates were analyzed by LC-MS/MS. Owing to the enrichment provided by specific probes, activity-based protein profiling...
### Table 1. Platelet Membrane Receptors

<table>
<thead>
<tr>
<th>Uniprot</th>
<th>Gene</th>
<th>Protein Name</th>
<th>Burkhart et al&lt;sup&gt;42&lt;/sup&gt;</th>
<th>Wijten et al&lt;sup&gt;97&lt;/sup&gt;</th>
</tr>
</thead>
<tbody>
<tr>
<td>Q9H7M9</td>
<td>C10orf54</td>
<td>Platelet receptor Gi24</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q96EK7</td>
<td>CPG</td>
<td>Constitutive coactivator of PPAR-γ</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P51679</td>
<td>CCR4</td>
<td>C-C chemokine receptor type 4</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q6YHK3</td>
<td>CD109</td>
<td>CD109 antigen</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P48509</td>
<td>CD151</td>
<td>CD151 antigen</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q15782</td>
<td>CD226</td>
<td>CD226 antigen</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P29965</td>
<td>CD40L</td>
<td>CD40 ligand</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q08722</td>
<td>CD47</td>
<td>Leukocyte surface antigen CD47</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P13987</td>
<td>CD59</td>
<td>CD59 antigen</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P08962</td>
<td>CD63</td>
<td>CD63 antigen</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P27701</td>
<td>CD82</td>
<td>CD82 antigen</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P21926</td>
<td>CD9</td>
<td>CD9 antigen</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P61073</td>
<td>CXCR4</td>
<td>C-X-C chemokine receptor type 4</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P98172</td>
<td>EFNB1</td>
<td>Ephrin-B1</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q96AP7</td>
<td>ESAM</td>
<td>Endothelial cell–selective adhesion molecule</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P30273</td>
<td>FCER1G</td>
<td>High-affinity immunoglobulin epsilon receptor subunit γ</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P12318</td>
<td>FCGR2A</td>
<td>Low-affinity immunoglobulin γ</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P13224</td>
<td>GP1BB</td>
<td>Platelet glycoprotein Ib β chain</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P16671</td>
<td>CD36</td>
<td>Platelet glycoprotein IV</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P40197</td>
<td>GP5</td>
<td>Platelet glycoprotein V</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q9HCN6</td>
<td>GP6</td>
<td>Platelet glycoprotein VI</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P14770</td>
<td>GP9</td>
<td>Platelet glycoprotein IX</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P13598</td>
<td>ICAM2</td>
<td>Intercellular adhesion molecule 2</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P08514</td>
<td>ITGA2B</td>
<td>Platelet glycoprotein lb</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P05106</td>
<td>ITGB3</td>
<td>Platelet glycoprotein III</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q09624</td>
<td>JAM1</td>
<td>Junctional adhesion molecule A</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q98X67</td>
<td>JAM3</td>
<td>Junctional adhesion molecule C</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>O06042</td>
<td>NRP2</td>
<td>Neutropin-2</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P06756</td>
<td>P06756L</td>
<td>Integrin α-V, CD51</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P51575</td>
<td>P2RX1</td>
<td>P2RX1 purinoceptor</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P25116</td>
<td>PAR1</td>
<td>Proteinase-activated receptor 1</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q96R0</td>
<td>PAR4</td>
<td>Proteinase-activated receptor 4</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q5V43</td>
<td>PEAR1</td>
<td>Platelet endothelial aggregation receptor 1</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P16284</td>
<td>PECAM1</td>
<td>Platelet endothelial cell adhesion molecule 1</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q09CM2</td>
<td>PLXNA4</td>
<td>Plexin-A4</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q15031</td>
<td>PLXNB2</td>
<td>Plexin-B2</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q0UL4</td>
<td>PLXNB3</td>
<td>Plexin-B3</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P43119</td>
<td>PTGIR</td>
<td>Prostacyclin receptor</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P16109</td>
<td>SELP</td>
<td>P-selectin</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>O92854</td>
<td>SEMA4D</td>
<td>Semaphorin-4D</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P21731</td>
<td>TBX2AR</td>
<td>Thromboxane A2 receptor</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q92956</td>
<td>TNFRSF14</td>
<td>Tumor necrosis factor receptor superfamily member 14</td>
<td>+</td>
<td>+</td>
</tr>
</tbody>
</table>

(continued)

<table>
<thead>
<tr>
<th>Uniprot</th>
<th>Gene</th>
<th>Protein Name</th>
<th>Burkhart et al&lt;sup&gt;42&lt;/sup&gt;</th>
<th>Wijten et al&lt;sup&gt;97&lt;/sup&gt;</th>
</tr>
</thead>
<tbody>
<tr>
<td>P25942</td>
<td>CD40</td>
<td>Tumor necrosis factor receptor superfamily member 5</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q15628</td>
<td>TRADD</td>
<td>Tumor necrosis factor receptor type 1</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P40238</td>
<td>TPOR</td>
<td>Thrombopoietin receptor</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P37288</td>
<td>AVPR1A</td>
<td>Vasopressin V1a receptor</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q9NYP3</td>
<td>CD93</td>
<td>Complement component C1q receptor</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P06213</td>
<td>INSR</td>
<td>Insulin receptor</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q9H1C0</td>
<td>LPAR5</td>
<td>Lyso phosphatidic acid receptor 5</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>O1590</td>
<td>P2RY1</td>
<td>P2RY1 purinoceptor</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>O9H244</td>
<td>P2RY12</td>
<td>P2RY12 purinoceptor</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P51805</td>
<td>PLXNA3</td>
<td>Plexin-A3</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>O06486</td>
<td>PLXNC1</td>
<td>Plexin-C1</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>P25105</td>
<td>PTAFR</td>
<td>Platelet-activating factor receptor</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>Q13258</td>
<td>PTGDR</td>
<td>Prostaglandin D2 receptor</td>
<td>+</td>
<td>+</td>
</tr>
<tr>
<td>O75888</td>
<td>TNFSF13</td>
<td>Tumor necrosis factor ligand superfamily member 13</td>
<td>−</td>
<td>+</td>
</tr>
<tr>
<td>P29274</td>
<td>ADORA2A</td>
<td>Adenosine receptor A2a</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P32246</td>
<td>CCR1</td>
<td>C-C chemokine receptor type 1</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P51677</td>
<td>CCR1</td>
<td>C-C chemokine receptor type 3</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P08174</td>
<td>CD55</td>
<td>CD55 antigen</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P21854</td>
<td>CD72</td>
<td>B-cell differentiation antigen CD72</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P49238</td>
<td>CX3CR1</td>
<td>CX3C chemokine receptor 1</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P35462</td>
<td>DRD3</td>
<td>Dopamine D3 receptor</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P21918</td>
<td>DRD5</td>
<td>Dopamine D5 receptor</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P52798</td>
<td>EFN4</td>
<td>Ephrin-A4</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P28223</td>
<td>HTR2A</td>
<td>Serotonin receptor 5-HT2A</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P48357</td>
<td>LEPR</td>
<td>Leptin receptor</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>O14786</td>
<td>NRPI</td>
<td>Neutropin-1</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P16234</td>
<td>PDGFRα</td>
<td>Platelet-derived growth factor receptor α</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P09619</td>
<td>PDGFRB</td>
<td>Platelet-derived growth factor receptor β</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>Q14242</td>
<td>SELPLG</td>
<td>P-selectin glycoprotein ligand 1</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td>P35590</td>
<td>TIE1</td>
<td>Tyrosine protein kinase receptor Tie-1</td>
<td>−</td>
<td>−</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Prostaglandin E2 receptors</td>
<td>−</td>
<td>−</td>
</tr>
</tbody>
</table>

From a set of 73 known platelet membrane receptors,125 ≈75% have been identified by the 2 most comprehensive global proteome analyses. Notably, this list might contain nonplatelet proteins (e.g., as the insulin receptor could not be detected immunologically),126 can be highly sensitive and access proteins hidden from common proteome analyses. However, it should be kept in mind that they can also enrich nonplatelet contaminations.

**PTM and Signaling Proteomics in Platelets**

PTM have a strong effect on protein structure and, therefore, can modulate protein function, localization, interaction, and
What Can Quantitative Proteomics Tell Us About Platelet Biology?

Because the proteomics field is rapidly moving from discovery/inventory (listing the proteins and PTM in a cell) to understanding/function (knowing the numbers of proteins and their dynamics), it becomes relevant to consider what proteomics can teach us about the precise functions of cells in health and disease. Our first large-scale quantitative analysis of the human platelet proteome, covering ≈80% of the entire amount of proteins in platelets, has refined the insight into the physiological functions of platelets in several ways. Figure 4 provides an overview of the cumulative copy numbers of the proteins and 11 protein kinase G substrates have been confirmed.

Lately, a new PKA and protein kinase G substrate was identified, namely calcium and diacylglycerol-regulated guanine nucleotide exchange factor I, the main activator for Rap1β in platelets. Quantitative MS analysis of 4 putative phosphorylation sites revealed Ser387 as the major PKA target, which on phosphorylation causes inhibition of Rap1β activity.

To gain an improved insight into the inhibitory cAMP/PKA pathway, we used quantitative MS and monitored time-resolved changes of phosphorylation patterns in human platelets treated with different concentrations of the stable prostacyclin analog, iloprost. Quantifying >2700 phosphorylation sites over 4 different time points, we could identify =300 unique proteins that are differentially phosphorylated on platelet inhibition. Our data indicate that platelet inhibition is not merely mediated by PKA but involves the interaction with other signaling pathways. For instance, 16 protein kinases and 7 protein phosphatases showed differential phosphorylation patterns on iloprost treatment, including increased phosphorylation of the inhibitory site Tyr530 of Src kinase.

Beside phosphorylation, glycosylation is among the most abundant PTM, especially for extracellular proteins, as such cell surface markers and secretory proteins. Glycosylation can either occur as a static modification or as a dynamic signaling process. Even with state-of-the-art instrumentation, the extreme heterogeneity and complexity of glycosylation patterns render their analysis challenging, such that many studies focus on the enrichment and identification of glycosylated residues rather than on characterizing the glycan structures. Yet, glycosylation patterns may dramatically influence processes such as platelet biogenesis and activation. To date, only few dedicated glycoproteomics studies have been conducted on human platelets, identifying several hundreds of N-glycosylation sites mainly on membrane proteins. However, the advent of more sensitive and robust protocols for the enrichment and quantitative analysis of glycopeptides is promising for the future.

Another PTM in platelets that was recently studied using proteomics is palmitoylation. Palmitoylation is the covalent but yet reversible attachment of palmitate to cysteine residues, thereby modulating protein–protein and protein–lipid interactions. Using a dedicated enrichment protocol in conjunction with MS, Dowal et al identified 215 palmitoylated proteins. However, the authors did not directly detect the intact PTM by MS, and additional work is required to confirm these findings.

What Can Quantitative Proteomics Tell Us About Platelets?

Because the proteomics field is rapidly moving from discovery/inventory (listing the proteins and PTM in a cell) to understanding/function (knowing the numbers of proteins and their dynamics), it becomes relevant to consider what proteomics can teach us about the precise functions of cells in health and disease. Our first large-scale quantitative analysis of the human platelet proteome, covering ≈80% of the entire amount of proteins in platelets, has refined the insight into the physiological functions of platelets in several ways. Figure 4 provides an overview of the cumulative copy numbers of the proteins and 11 protein kinase G substrates have been confirmed.

Lately, a new PKA and protein kinase G substrate was identified, namely calcium and diacylglycerol-regulated guanine nucleotide exchange factor I, the main activator for Rap1β in platelets. Quantitative MS analysis of 4 putative phosphorylation sites revealed Ser387 as the major PKA target, which on phosphorylation causes inhibition of Rap1β activity.

To gain an improved insight into the inhibitory cAMP/PKA pathway, we used quantitative MS and monitored time-resolved changes of phosphorylation patterns in human platelets treated with different concentrations of the stable prostacyclin analog, iloprost. Quantifying >2700 phosphorylation sites over 4 different time points, we could identify =300 unique proteins that are differentially phosphorylated on platelet inhibition. Our data indicate that platelet inhibition is not merely mediated by PKA but involves the interaction with other signaling pathways. For instance, 16 protein kinases and 7 protein phosphatases showed differential phosphorylation patterns on iloprost treatment, including increased phosphorylation of the inhibitory site Tyr530 of Src kinase.

Beside phosphorylation, glycosylation is among the most abundant PTM, especially for extracellular proteins, as such cell surface markers and secretory proteins. Glycosylation can either occur as a static modification or as a dynamic signaling process. Even with state-of-the-art instrumentation, the extreme heterogeneity and complexity of glycosylation patterns render their analysis challenging, such that many studies focus on the enrichment and identification of glycosylated residues rather than on characterizing the glycan structures. Yet, glycosylation patterns may dramatically influence processes such as platelet biogenesis and activation. To date, only few dedicated glycoproteomics studies have been conducted on human platelets, identifying several hundreds of N-glycosylation sites mainly on membrane proteins. However, the advent of more sensitive and robust protocols for the enrichment and quantitative analysis of glycopeptides is promising for the future.

Another PTM in platelets that was recently studied using proteomics is palmitoylation. Palmitoylation is the covalent but yet reversible attachment of palmitate to cysteine residues, thereby modulating protein–protein and protein–lipid interactions. Using a dedicated enrichment protocol in conjunction with MS, Dowal et al identified 215 palmitoylated proteins. However, the authors did not directly detect the intact PTM by MS, and additional work is required to confirm these findings.

What Can Quantitative Proteomics Tell Us About Platelet Biology?

Because the proteomics field is rapidly moving from discovery/inventory (listing the proteins and PTM in a cell) to understanding/function (knowing the numbers of proteins and their dynamics), it becomes relevant to consider what proteomics can teach us about the precise functions of cells in health and disease. Our first large-scale quantitative analysis of the human platelet proteome, covering ≈80% of the entire amount of proteins in platelets, has refined the insight into the physiological functions of platelets in several ways. Figure 4 provides an overview of the cumulative copy numbers of the proteins and 11 protein kinase G substrates have been confirmed.

Lately, a new PKA and protein kinase G substrate was identified, namely calcium and diacylglycerol-regulated guanine nucleotide exchange factor I, the main activator for Rap1β in platelets. Quantitative MS analysis of 4 putative phosphorylation sites revealed Ser387 as the major PKA target, which on phosphorylation causes inhibition of Rap1β activity.

To gain an improved insight into the inhibitory cAMP/PKA pathway, we used quantitative MS and monitored time-resolved changes of phosphorylation patterns in human platelets treated with different concentrations of the stable prostacyclin analog, iloprost. Quantifying >2700 phosphorylation sites over 4 different time points, we could identify =300 unique proteins that are differentially phosphorylated on platelet inhibition. Our data indicate that platelet inhibition is not merely mediated by PKA but involves the interaction with other signaling pathways. For instance, 16 protein kinases and 7 protein phosphatases showed differential phosphorylation patterns on iloprost treatment, including increased phosphorylation of the inhibitory site Tyr530 of Src kinase.

Beside phosphorylation, glycosylation is among the most abundant PTM, especially for extracellular proteins, as such cell surface markers and secretory proteins. Glycosylation can either occur as a static modification or as a dynamic signaling process. Even with state-of-the-art instrumentation, the extreme heterogeneity and complexity of glycosylation patterns render their analysis challenging, such that many studies focus on the enrichment and identification of glycosylated residues rather than on characterizing the glycan structures. Yet, glycosylation patterns may dramatically influence processes such as platelet biogenesis and activation. To date, only few dedicated glycoproteomics studies have been conducted on human platelets, identifying several hundreds of N-glycosylation sites mainly on membrane proteins. However, the advent of more sensitive and robust protocols for the enrichment and quantitative analysis of glycopeptides is promising for the future.

Another PTM in platelets that was recently studied using proteomics is palmitoylation. Palmitoylation is the covalent but yet reversible attachment of palmitate to cysteine residues, thereby modulating protein–protein and protein–lipid interactions. Using a dedicated enrichment protocol in conjunction with MS, Dowal et al identified 215 palmitoylated proteins. However, the authors did not directly detect the intact PTM by MS, and additional work is required to confirm these findings.
The relatively high abundance of proteins that can be classified as cytoskeleton-linked signaling proteins (9%) is also of particular interest. Most prominent among these are the integrin α_{IIb}β_{3}-linked proteins, talin (116k), fermitin member URP2 (116k), filamin A (88k), and vinculin (81k). Linked to the actin cytoskeleton via GPIb-V-IX complexes (von Willebrand factor receptors) are the so-called 14-3-3 adapter or modifier proteins, of which 5 members seem to be highly expressed (sum 572k). The cytoskeleton also plays an important role in regulating α_{IIb}β_{3}, and GPIb-V-IX as 2 key adhesive receptors, and data from proteomics have added to this recognition by revealing actin cytoskeleton rearrangements connected to these receptors. Moreover, both α_{IIb} (64k–83k) and GPIb-V-IX (49k) are present in the top-100 platelet proteins, thus providing sufficient copy numbers to modulate actin cytoskeleton-dependent platelet shape. Interestingly, the only other highly expressed membrane protein is the chloride channel CLIC1 (44k), the function of which is still poorly understood, but may include modulation of ADP receptor signaling.

Among the most profusely expressed cytosolic proteins are small GTP-binding proteins and their regulators, in particular 2 variants of Rap1b (299k) and RapA (125k), which are known to regulate integrin activation in a cytoskeleton-dependent way. Other abundant small GTP-binding proteins are 2 ADP-ribosylation factors (80k) and Rab27b (36k), of which at least the latter controls platelet dense granule secretion. Directly or indirectly linked to actin-dependent integrin signaling are also the adaptor protein, VASP (45k), and the integrin-linked kinase (60k). The same holds for the Ca^{2+}-dependent protease, calpain-I (51k), which cleaves many cytoskeletal components and controls integrin closure in procoagulant platelets.

Another remarkable fact is that 2 of the highly expressed signaling proteins play a role in PKA-dependent inhibition of platelets, that is, the adenyl cyclase–associated protein CAP1 (42k) and VASP, which is the major substrate of this cAMP-dependent protein kinase. Although earlier data with mice indicated that the prostacyclin receptor (a key cAMP-elevating receptor on platelets) is not required for normal hemostasis, the abundance of these proteins supports a quantitatively important role of cAMP signaling in the control of platelet functions (see also above). Interestingly, the majority of top-100 proteins controlling metabolic processes are confined to glucose metabolism (glyceraldehyde 3-phosphate dehydrogenase, fructose bisphosphate aldolase, α/γ-enolase, pyruvate kinase, lactate dehydrogenase, triosephosphate isomerase, and phosphoglycerate kinase; each 36k–106k). Hence, glycolysis seems to be most essential for the metabolic control of platelet functionality.

When comparing the categories of top-100 proteins with those of the top-200 (13 929k copies, >16k per protein) and top-500 (16 943k copies, >6.7k per protein), several changes are noteworthy (Figure 4). The top-200 and top-500 sets show a small decrease in the relative abundance of microtubule and actin–myosin cytoskeletal proteins, as well as in α-granular proteins, thus indicating that the most prominent of these proteins are expressed at high copy numbers. The same holds for the category of cytoskeletal-linked signaling proteins. In
contrast, there is an increase in abundance of small GTPases and their regulators (5%–8%), with major contributions of the (regulators of) Cdc42, Rab, Rac, and Rho isoforms that organize intracellular membrane interactions and platelet shape change. Furthermore, the top-500 comprises a marked increase in copy numbers of signaling and adaptor proteins (6%–12%). These include significant numbers of the most poorly defined adapter proteins; several proteins of the ubiquitin–proteasome system (sum >80k); well-known (Src, Btk, Lyn) and less known tyrosine protein kinases; various serine/threonine protein kinases (protein kinase Cα, PKA, ROCK, MAPK isoforms); protein phosphatases (G6b, PP1-3 isoforms); Ca²⁺-binding proteins; classical GTP-binding proteins Grq and Gq; and components of the thromboxane synthase complex. Prominent membrane receptor proteins in the top-500 are several HLA-class histocompatibility antigens, the collagen receptor GPVI, the tetraspanin CD9, and the thrombospondin receptor CD36. Jointly, the majority of these proteins can be seen as part of the hard-core signaling network, controlling platelet activation to adhesiveness, secretion, and aggregation. As described above, quantitative analysis of the most abundant (structural) proteins can provide general information on cellular functions. However, for proteins that are more scarcely expressed, the link to function is less obvious because often the extent and type of PTM may be even more important for the protein activity than the expression number. An example is provided by the platelet protein kinase C-α/β isoforms (8k–10k) that regulate almost every platelet response. These protein kinases not only require multiple phosphorylation steps to become catalytically active, but their activity is also modulated by additional phosphorylation events on platelet activation. Hence, for full understanding of the role of the majority of cellular proteins, quantitative information about phosphorylation processes and other PTM is needed. Based on the literature, >2500 phosphorylation sites and >1000 other PTM have been identified in platelets. However, this information has not been coupled to the quantitative proteome. To date, there is limited evidence that the platelet protein composition may vary with age and in patients with cardiovascular or other diseases. In addition, there are still groups of proteins that are difficult to access using proteomics but are nevertheless important for platelet function. Accordingly, current global proteomics procedures might result in underestimation of heavily modified and membrane proteins with high hydrophobicity; splice variants, isoforms of related proteins, and proteins expressed at low copy numbers (<0.5k). More dedicated approaches using enrichment procedures or targeted MS analyses described above need to be used in these cases.

Current and Future Use of Proteomics to Study Platelet Function Disorders

It is to be expected that many functional disorders of platelets are attributable to alterations in protein expression levels and PTM patterns, which might be reflected already in the composition of the top-100/200/500 proteins derived from the quantitative proteome. Despite the tremendous efforts made in the past 130 years of platelet research, even for well-established diseases such as Glanzmann thrombasthenia, we know little more than the single protein defect. However, further secondary changes on the protein and PTM level can be expected. This is in accordance with our own preliminary quantitative proteome data on platelets from healthy donors treated with aspirin or serotonin uptake inhibitors, as well as platelets from thrombocytopenic patients. Already from a limited number of analyses, unanticipated proteins show differential expression patterns and might represent target proteins for follow-up investigations.

The most thoroughly characterized platelet disorders are those associated with an impairment in platelet function or too few or too many platelets (ie, thrombocytopenia and thrombocytosis, respectively). An increase in platelet count above the normal range (150–400×10⁶/mL) is known as essential thrombocythaemia. This is a rare chronic disorder that is associated with the overproduction of platelets in the bone marrow. Approximately 50% of patients have a V617F mutation in the tyrosine kinase JAK2, which gives rise to a low level of constitutive signaling and potentiation of the response to the thrombopoietin receptor, c-Mpl. The phenotype is influenced by other modifier genes and is associated with small
changes in expression of surface glycoprotein receptors in either direction. The diagnosis of essential thrombocythaemia is achieved by ruling out other causes of an increase in platelet count and a genetic test for V617F. The causative mutation in V617F-negative essential thrombocythaemia patients is not known, and proteomics may have a role in their investigation. This potential, however, needs to take into account the heterogeneity in protein levels between patients and the potential influence of platelet-lowering drugs such as anagrelide.

Thrombocytopenia can be inherited or acquired, with the most common form of the conditions being through immune thrombocytopenia. In this acquired, antibody-mediated disorder, the use of proteomics is needed to understand why some but not all patients experience severe bleeding episodes, despite having similar platelet counts. Proteomics may also help in the diagnosis of immune thrombocytopenia. The demonstration of a platelet proteome that is different to that of healthy controls may provide evidence of a contributing genetic cause of the reduced platelet count.

Individuals with inherited thrombocytopenia are in comparison relatively rare, numbering in the order of 1:100000 of the population, with many of these having a mild reduction in platelet count of <40% of the normal range. Many individuals with a mild reduction in platelet have no clinical history of excessive bleeding and are diagnosed through routine blood tests (eg, in association with pregnancy). However, other individuals with a similar, mild reduction in platelet count have a significant history of bleeding, indicative of impairment of platelet function. In ≈50% of cases, the causative gene is known, ranging from transcription factors (eg, Runx1 and GATA-1) to proteins that play a critical role in platelet formation (eg, GP Ibα, MYH9, and WASP). Proteomics has the potential to find causative genes (proteins) in individuals, in whom the genetic defect has not yet been identified, either through the identification of the gene or as a guide to next-generation sequencing studies.

The most heterogeneous and challenging group of patients for investigation are those with a clinical history of bleeding consistent with an impairment in platelet activation but where the cause is not known. Such patients are relatively rare, in the order of 1:10000 in the population. In approximately half of these, a defect in platelet function can be demonstrated using the gold standard platelet function assay of Born aggregometry. Many of these patients present early in life with bleeding symptoms, suggesting that the defect is inherited, but a significant number only present later on when subject to a challenge such as surgery, tooth removal, or menstruation. In patients from nonconsanguineous relationships, the major challenge is the lack of penetrance between family members, reflecting the multifactorial nature of the bleeding diathesis. In other words, the affected individual has almost certainly inherited >1 mutation, and it is the combination of mutations that gives rise to bleeding. Although next-generation sequencing approaches are being increasingly used to identify candidate mutations, the demonstration of cause and effect will be dependent on platelet functional studies and techniques, such as quantitative proteomics, to demonstrate significant changes in protein expression or function. Even then, it will be challenging to establish the degree to which bleeding is dependent on the gene mutations and the corresponding changes in platelet function.

The challenge in linking candidate mutations in platelets and changes in protein expression is even higher in disorders involving platelet activation including arterial thrombosis. Platelets are now recognized for their important roles in inflammation, cancer metastasis, innate defense, organ development, and repair, as well as in an increasing number of cardiovascular diseases, including deep vein thrombosis, atherosclerosis, tissue infarctions, and ischemia–reperfusion injury. The molecular basis of the role of platelets in the majority of these pathologies is still under debate, thereby making it challenging to use proteomics to establish the molecular mechanisms through which platelets contribute to disease processes.

### Future Directions of Platelet Proteomics

In the past few years, it has become increasingly clear that platelet proteomics can provide novel insights into basic research questions and thus improve our understanding about the fundamental processes that regulate platelets and can also contribute to the diagnosis of platelet disorders, as summarized in Table 2. The protein composition and the phosphorylation patterns of platelets will be useful to understand certain disease states and therapeutic interventions. In particular, quantitative phosphoproteomic studies will pave the way for a refined understanding of platelet properties, beyond the classical biochemical studies which focused on the description of linear pathways such as stimulus→receptor→second messenger→phosphoprotein→effect on cell function. Although these approaches undeniably have made and will make important discoveries, the past few years clearly showed

### Table 2. Future Applications of Quantitative Proteomics and Phosphoproteomics

<table>
<thead>
<tr>
<th>Basic Research Questions</th>
<th>Diagnosis of Platelet Disorders</th>
</tr>
</thead>
<tbody>
<tr>
<td>Comparison of protein composition</td>
<td>Comparison of protein composition</td>
</tr>
<tr>
<td>Healthy individuals: patients with known hereditary platelet or blood cell disorder</td>
<td>Healthy individuals: patients with established platelet dysfunction of unknown origin</td>
</tr>
<tr>
<td>➜ Understanding the full phenotype and pathomechanism of hereditary platelet and blood cell disorders</td>
<td>➜ Discovering and understanding novel platelet defects</td>
</tr>
<tr>
<td>Comparison of protein composition</td>
<td>Comparison of protein composition</td>
</tr>
<tr>
<td>Healthy individuals: patients treated with antiplatelet drugs</td>
<td>Healthy individuals: patients with chronic diseases (coagulation disorders, vascular, inflammatory, metabolic diseases)</td>
</tr>
<tr>
<td>➜ Understanding effects of chronic treatment on platelet protein composition and function</td>
<td>➜ Determining the signature of chronic diseases</td>
</tr>
<tr>
<td>Comparison of phosphoproteome</td>
<td>Comparison of phosphoproteome (activatory or inhibitory pathways)</td>
</tr>
<tr>
<td>Healthy individuals: platelets treated with agonists or inhibitors</td>
<td>Healthy individuals: patients with chronic coagulation disorders or vascular, inflammatory, metabolic disease</td>
</tr>
<tr>
<td>➜ Defining the complex signaling pathways in platelets</td>
<td>➜ Understanding the pathogenesis of the diseases</td>
</tr>
</tbody>
</table>
that signaling is much more complicated, branched, and non-linear than originally anticipated, including a considerable level of cross talk, in platelets for instance between inhibitory and activatory pathways. In this context, the extensive cross talk between different PTM and the existence of a PTM code that defines the functions of proteins, protein complexes, and signaling pathways have gained more and more attention. The vast amount of data that comes from quantitative proteomics studies, however, will require a thorough evaluation using systems biology approaches not only for obtaining a systematic interpretation of signaling networks and nodes but also for the simulation of experiments and conditions that cannot be realized in vivo or in vitro. This will lead to the description of (phospho)protein clusters that are characteristic for the various levels of inhibitory and activation pathways or that represent molecular signatures for those pathways, which are known to be affected in disease. We expect that recent strategies to analyze protein complexes using quantitative MS will further improve our understanding of such molecular clusters. In addition, quantitative MS-based profiling methods for cell organelles can provide information about the spatial distribution of proteins and nodes but also for the simulation of experiments and conditions that cannot be realized in vivo or in vitro. This will lead to the description of (phospho)protein clusters that are characteristic for the various levels of inhibitory and activation pathways or that represent molecular signatures for those pathways, which are known to be affected in disease.

Acknowledgments
We thank Florian Beck, Jörg Geiger, Craig Hughes, Urs Lewandowski, Stefan Loroch, Nadine Mattheij, Jan Moebius, Fiorella A. Solari, Frauke Swieringa, and Marc Vaudel for their valuable contribution to former and present platelet proteomics projects.

Sources of Funding
The work received funding support from the Center for Translational Molecular Medicine (INCOAG) to J.W.M. Heemskerk; the Ministry of Innovation, Science and Research of the Federal State of Northrhine-Westphalia to J.M. Burkhart (BioNRW project z0911bt021a), A. Sickmann (BioNRW project z0911bt021a), and R.P. Zahedi; the Federal Ministry of Education and Research (BMBF) to J.M. Burkhart, A. Sickmann (31P5800), R.P. Zahedi, K. Jürk (01E01003), and U. Walter (01E01003); and the German Research Foundation (DFG, SFB668 TPA2) to S. Gambaryan, A. Sickmann, and U. Walter. S.P. Watson is a British Heart Foundation Chair (CH/03/003) and is also supported by the Wellcome Trust (088410).

Disclosures
None.

References


The complex transcriptional landscape of the anucleate human platelet.


human platelets isolated by two-dimensional gel electrophoresis by elec-
trospray ionization-tandem mass spectrometry (ESI-MS/MS) and liquid
chromatography-electrospray ionization-mass spectrometry (LC-ESI-
138. Parguña AF, Fernández Parguña, Grigorian-Shamajian L, Agra RM, 
Tejeira-Fernández E, Rosa I, Alonso J, Vítueloa-Roldán JE, Saez A, 
González-Juanatey JR, García A. Proteins involved in platelet signal-
ing are differentially regulated in acute coronary syndrome: a proteomic 
139. Senis YA, Antrobus R, Severin S, Parguña AF, Rosa I, Zitzmann N, 
Watson SP, García A. Proteomic analysis of integrin alphabeta3
outside-in signaling reveals Src-kinase-independent phosphorylation of 
Dok-1 and Dok-3 leading to SHIP-1 interactions. *J Thromb Haemost.* 
140. Gegenbauer K, Elia G, Blanco-Fernández A, Smolenski A. Regulator of 
G-protein signaling 18 integrates activating and inhibitory signaling in 
of thrombin-activated platelets after in vivo nitric oxide-donor treatment: 
coordinated inhibition of phosphorylation (phosphatidylinositol 3-kinase-γ, 
14-3-3c, and growth factor receptor-bound protein 2) and cytoskeleton 
142. Smolenski A. Novel roles of cAMP/G protein-dependent signaling in plate-
Phosphorylation of Ca2DAG-GEFI by protein kinase A prevents Rap1 
Martens L, Walter U, Sickmann A. Time-resolved charac-
terization of cAMP/PKCa-dependent signaling reveals that platelet inhibi-
tion is a concerted process involving multiple signaling pathways. *Blood.* 
2014;123:e1–e10.
145. Apweiler R, Hermjakob H, Sharon N. On the frequency of protein gly-
cosylation, as deduced from analysis of the SWISS-PROT database. 
146. Wang Y, Jobe SM, Ding X, Choo H, Archer DR, Mi R, Ju T, Cummings 
RD. Platelet biogenesis and functions require correct protein O-glycosylation. 
147. Nowak AA, Canis K, Riddell A, Laffan MA, McKinnon TA. O-linked 
glycosylation of von Willebrand factor modulates the interaction with 
platelet receptor glycoprotein Ib under static and shear stress conditions. 
Glycosylation site analysis of human platelets by electrostatic repulsion 
hydrophilic interaction chromatography. *Clinical Proteomics.* 
149. Palmisano G, Parker BL, Engholm-Keller K, Lendal SE, Kalej S, 
Schult M, Schwümmle V, Graham ME, Saxtorph H, Cordwell SJ, Larsen MR. 
A novel method for the simultaneous enrichment, identification, and 
quantification of phosphoproteins and sialylated glycoproteins applied to 
150. Palmisano G, Lendal SE, Larsen MR. Titanium dioxide enrich-
ment of sialic acid-containing glycoproteins. *Methods Mol Biol.* 
2011;753:309–322.
151. Palmisano G, Lendal SE, Engholm-Keller K, Leth-Larsen R, Parker BL, 
Larsen MR. Selective enrichment of sialic acid-containing glycoproteins 
152. Heemskerk JW, Kaipers MJ, Munnicx CI, Siljander PR. Platelet col-


What Can Proteomics Tell Us About Platelets?
Julia M. Burkhart, Stepan Gambaryan, Stephen P. Watson, Kerstin Jurk, Ulrich Walter, Albert Sickmann, Johan W.M. Heemskerk and René P. Zahedi

Circ Res. 2014;114:1204-1219
doi: 10.1161/CIRCRESAHA.114.301598

Circulation Research is published by the American Heart Association, 7272 Greenville Avenue, Dallas, TX 75231
Copyright © 2014 American Heart Association, Inc. All rights reserved.
Print ISSN: 0009-7330. Online ISSN: 1524-4571

The online version of this article, along with updated information and services, is located on the World Wide Web at:
http://circres.ahajournals.org/content/114/7/1204

Permissions: Requests for permissions to reproduce figures, tables, or portions of articles originally published in Circulation Research can be obtained via RightsLink, a service of the Copyright Clearance Center, not the Editorial Office. Once the online version of the published article for which permission is being requested is located, click Request Permissions in the middle column of the Web page under Services. Further information about this process is available in the Permissions and Rights Question and Answer document.

Reprints: Information about reprints can be found online at:
http://www.lww.com/reprints

Subscriptions: Information about subscribing to Circulation Research is online at:
http://circres.ahajournals.org/subscriptions/