

OPEN

Received: 27 September 2018 Accepted: 16 April 2019 Published online: 21 May 2019

Hair Proteome Variation at Different Body Locations on Genetically Variant Peptide Detection for Protein-Based Human Identification

Fanny Chu^{1,2}, Katelyn E. Mason¹, Deon S. Anex¹, A. Daniel Jones^{2,3} & Bradley R. Hart¹

Human hair contains minimal intact nuclear DNA for human identification in forensic and archaeological applications. In contrast, proteins offer a pathway to exploit hair evidence for human identification owing to their persistence, abundance, and derivation from DNA. Individualizing single nucleotide polymorphisms (SNPs) are often conserved as single amino acid polymorphisms in genetically variant peptides (GVPs). Detection of GVP markers in the hair proteome via high-resolution tandem mass spectrometry permits inference of SNPs with known statistical probabilities. To adopt this approach for forensic investigations, hair proteomic variation and its effects on GVP identification must first be characterized. This research aimed to assess variation in single-inch head, arm, and pubic hair, and discover body location-invariant GVP markers to distinguish individuals. Comparison of protein profiles revealed greater body location-specific variation in keratin-associated proteins and intracellular proteins, allowing body location differentiation. However, robust GVP markers derive primarily from keratins that do not exhibit body location-specific differential expression, supporting GVP identification independence from hair proteomic variation at the various body locations. Further, pairwise comparisons of GVP profiles with 8 SNPs demonstrated greatest interindividual variation and high intraindividual consistency, enabling similar differentiative potential of individuals using single hairs irrespective of body location origin.

Human hair, as one of the few biological specimen types that persist for long periods of time, is invaluable to forensic and archaeological investigations, yet limited identification information is often obtained from hair using conventional approaches. Comprised primarily of keratins and keratin-associated proteins, hair exhibits high durability that contributes to its persistence. Packed into coiled-coils and localized to the cortex, cuticle, or medulla of the hair shaft 1,2 , hair keratins are stabilized and provide tensile strength via cross-linking by cystine disulfide bonds 3 . Differences in their amino acid composition further separate them into type I acidic keratins (K31, K32, K33A, K33B, K34 – K40) with low-sulfur content and type II neutral to basic keratins (K81 – K86) with high-sulfur content $^{2-6}$. Keratin-associated proteins (KAPs), categorized by their amino acid composition as high-sulfur, ultra-high sulfur, or high glycine-tyrosine proteins, participate in cross-linking alongside keratins to provide rigidity to the matrix 7,8 .

In contrast to its high protein content, hair contains minimal intact nuclear DNA, which is the preferred evidence type for human identification in forensic investigations owing to its specificity for individualization and low associated statistical probabilities. Found within keratinized corneocytes, nucleases including DNase1L2 degrade DNA to tiny fragments, depending on their abundances and catalytic activities. This process begins immediately as the keratinocyte moves out of the hair follicle and terminal differentiation occurs¹⁰, and exhibits high interindividual variation. With continued hair growth, DNA degrades to low and variable amounts, making forensic

¹Forensic Science Center, Lawrence Livermore National Laboratory, 7000 East Ave., Livermore, CA, 94550, USA. ²Department of Chemistry, Michigan State University, 578 S Shaw Ln, East Lansing, MI, 48824, USA. ³Department of Biochemistry and Molecular Biology, Michigan State University, 603 Wilson Rd, East Lansing, MI, 48824, USA. Correspondence and requests for materials should be addressed to D.S.A. (email: anex1@llnl.gov)

nuclear DNA profiling in hair unreliable. When complete analysis of nuclear DNA, the gold standard, either through short tandem repeat profiling $^{10-12}$ or with single nucleotide polymorphisms (SNPs) $^{13-15}$, is not possible, an alternative method for human identification becomes critically important.

Proteins offer an attractive pathway for identification, as they are robust and derive from DNA. Protein-based identification methods are largely supported by advancements in proteomics, such as faster-scanning high-resolution mass analyzers and bioinformatics tools. The proteome is a focus for biomarker identification indicative of various disease states and also for analysis of ancient specimens, since minimal DNA remains owing to the ubiquitous nature of nucleases in the environment^{16,17}, but proteins are likely to be preserved. High-resolution mass spectrometry analysis of fossil bones revealed survival of proteins in the leucine-rich repeat family and serum proteins in samples dating back to 900 thousand years¹⁸. Peptide markers characterized in various keratinous tissues of agricultural and archaeological importance, including horn and baleen, enabled species identification within the mammalian kingdom based on variant amino acid sequences that distinguish genera¹⁹. Proteomics methods for biological fluid identification have also identified peptide markers specific to fluids including human saliva, urine, seminal fluid, and vaginal fluid, demonstrating the potential of protein detection and identification for utility in forensic investigations^{20,21}. While the hair proteome has been probed to characterize disorders including lamellar ichthyosis and trichothiodystrophy using protein expression differences^{8,22,23}, hair proteins have received little attention for human identification.

The few reports of distinguishing individuals from hair proteomes include a 2014 paper by Laatsch and colleagues who analyzed protein composition and expression differences in hair from different ethnic populations and from various body locations and found specific protein expression differences to differentiate based upon ethnicity and body location²⁴. Following in this vein, Wu *et al.* showed in 2017 that protein profiles differentiated monozygotic twins from unrelated individuals²⁵. However, both studies focused on protein abundances to provide differentiation of different populations, which does not provide sufficient specificity to distinguish individuals, as hair protein levels are evolutionarily conserved within ethnic populations.

Alternatively, genetically variant peptides (GVPs) in hair proteins possess great potential for differentiation of individuals. Identification of single amino acid polymorphisms in GVPs permits inference of individualizing SNPs. Parker *et al.* first demonstrated the identification and use of single amino acid polymorphisms in GVPs from head hair to differentiate individuals²⁶. Further, analogous to a SNP panel developed by Pakstis *et al.*¹⁴, Parker and co-workers compiled a panel of 33 SNPs identified from GVP markers and verified by Sanger sequencing, and determined random match probabilities ranging up to 1 in 14,000 for a cohort of 60 subjects²⁶. Mason *et al.* then showed comparable performance in single one-inch hair analysis²⁷ to that using bulk quantities of scalp hair in Parker *et al.*²⁶, demonstrating the feasibility of protein-based human identification for forensic investigations. Although our protein-based identification approach has been adapted to simulate forensically-relevant sample sizes via single hair analysis, fundamental questions regarding hair protein chemistry and its effects on GVP marker identification remain.

One key knowledge gap lies in whether the same GVP markers can be identified in hair from different body locations, as hair origin is often not known from hair specimens recovered for forensic investigations. From work performed by Laatsch *et al.*, differential protein expression in hair from a few body locations has been characterized²⁴. However, effects of body location-specific protein abundance variation on robust identification of GVP markers have not yet been elucidated. We employ proteomics technologies and methodologies developed for single hair analysis in this study to examine GVP markers identified from head, arm, and pubic hair for any differences in the differentiative potential of individuals. Aims of this research include the determination of body location-specific proteomic variation, evaluation of the effects of differential protein expression on GVP identification and subsequent SNP inference, and quantification of the extent to which individuals are differentiated with robust, i.e., consistently identified and body location-invariant, GVP markers. We demonstrate the independence of forensic SNP identification from body location-specific hair proteomic variation and further identify viable GVP markers that yield similar distinction of individuals irrespective of body location origin in single one-inch (25 mm) hairs.

Results

Single inch hair sample preparation performance. Single one-inch hairs yield rich protein profiles that are comparable to profiles established with greater hair quantities; on average, 142 ± 33 (s.d.) proteins were identified from each of 9 head hairs (i.e., from three sets of proteomics-only biological replicates from three individuals), and the average number of identified unique peptides was $1,031 \pm 219$. From unique peptides, the average numbers of identified amino acids were $15,527 \pm 3,056$. The presence of a subset of unique peptides known as genetically variant peptides (GVPs) enabled inference of 16 ± 5 SNPs from major GVPs, and 17 ± 3 SNPs from minor GVPs (i.e., GVPs corresponding to the major and minor alleles, respectively). Because both major or minor GVPs allow SNP inference, non-synonymous, or missense, SNPs were reported for both types of GVPs. However, in some cases, detection of both GVPs for the same SNP may not be possible. In previous studies, Parker *et al.* identified at least 180 proteins in 10 mg of head hair samples from 60 subjects and detected between 156 and 2,011 unique peptides²⁶, and Adav *et al.* identified, on average, 195 ± 12 proteins in human hair using various sample preparation methods²⁸. Commensurate performance to previous works is achieved even when sample size is substantially reduced to simulate amounts of material available from forensic samples.

In addition, performing co-extraction of protein and mitochondrial DNA yielded no loss in protein information relative to processing for protein alone. Proteomic results from co-extraction were not statistically different from proteomics-only sample preparation for each of the above metrics (two sample t-test; $p \geq 0.106$; Supplementary Fig. S1); for example, 156 ± 56 proteins were identified from proteomics-only samples and 151 ± 39 proteins were detected in co-extracted samples. These observations indicate that additional steps taken to co-extract DNA with protein did not adversely affect protein identification or detection of unique peptides and

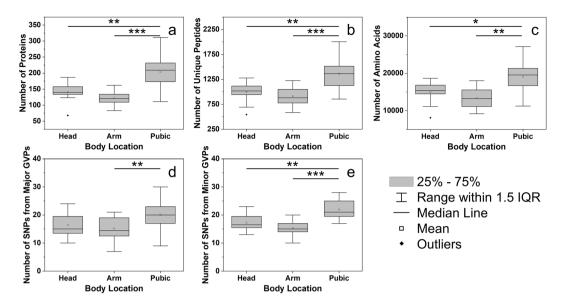


Figure 1. Comparison of numbers of identified (a) proteins, (b) unique peptides, (c) amino acids, and missense SNPs inferred from (d) major and (e) minor GVPs at different body locations. Black lines represent statistically significant comparisons and significance levels are represented as: $p \le 0.05$ (*), $p \le 0.01$ (**), and $p \le 0.001$ (***). Pubic hair samples yield statistically greater numbers of proteins, peptides, amino acids, and inferred SNPs (two-way ANOVA and Tukey HSD; n = 36).

missense SNPs from GVPs. As both sample preparation methods yielded the same proteomic information, the protein/DNA co-extracted sample set was included in this study for all further analyses. Analysis of GVPs and mtDNA can provide corroborating evidence for more confident profiling of individuals, which will be explored in a later publication.

Proteomic variation at different body locations. Hair proteomic variation at three different body locations in 36 hair specimens was first assessed by comparing five metrics: the numbers of detected proteins, unique peptides, amino acids, and missense SNPs from major and minor GVPs (Fig. 1). Two-way ANOVAs with Tukey HSD post-hoc tests were performed for each metric to account for effects of body location and individual. Statistical testing revealed significant effects of body location on the numbers of detected proteins ($p = 1.07 \times 10^{-4}$), unique peptides ($p = 5.66 \times 10^{-4}$), and amino acids ($p = 2.21 \times 10^{-3}$), while effects of individual and the interaction between body location and individual were not significant. A single inch of pubic hair yields more proteins, unique peptides, and amino acids, than head or arm hair. A significant effect of body location on the number of SNPs inferred from GVPs was observed for major ($p = 7.56 \times 10^{-3}$) and minor GVPs ($p = 1.91 \times 10^{-5}$). These results suggest that compared to head and arm hair, the protein composition of pubic hair is more complex, from which many GVPs and SNPs can be identified for human identification.

Significant effects of body location observed for these five metrics may arise from differences in mass per unit length of hair. The mass of a single inch of pubic hair (200.1 \pm 39.6 µg) was statistically greater than an inch of head (84.4 \pm 27.7 µg; two-way ANOVA and Tukey HSD; $p=1.76\times10^{-9}$) or arm hair (49.4 \pm 22.2 µg; $p=1.74\times10^{-11}$). Despite mass differences in hair, the same injection volume was used for each sample, and thus, different quantities of material were loaded onto the column for LC-MS/MS. It is proposed that more proteins, unique peptides, amino acids, and inferred SNPs were identified in pubic hair samples owing to larger on-column mass loadings.

To assess body location-specific proteomic variation without bias from different on-column mass loadings, protein abundances were examined after normalization to total chromatographic peak area of identified peptides. A previous study by Laatsch and co-workers reported differential expression at different body locations for a subset of proteins²⁴. To confirm these observations in this study for head and pubic hair, and to assess differential protein expression in arm hair, which was not examined previously, protein quantities derived from mass spectral data were compared. Various approaches have been utilized to quantify proteins using mass spectral data, including spectral counts^{24,29,30}, precursor ion peak areas from MS scans^{31,32}, and MS/MS fragment ion abundances³³ to represent peptide abundance. Because dynamic exclusion was used during data acquisition to maximize peptide identification and protein coverage, MS/MS spectral counts do not reliably represent peptide abundance, especially lower abundance peptides³⁴. We chose to use the more robust of the latter two methods and tabulated MS scan precursor ion peak areas in mass spectral data from a complete list of identified unique peptides. Bias towards samples with larger mass loadings was removed by normalizing each precursor ion peak area to the total peak area of all identified peptides. Protein abundance in each sample was calculated as the sum of all normalized peak areas assigned to the protein.

Protein abundance was examined in this study to observe any effects of body location. Statistical comparison of protein abundances identified 37 proteins with body location-specific differential expression, of which a subset

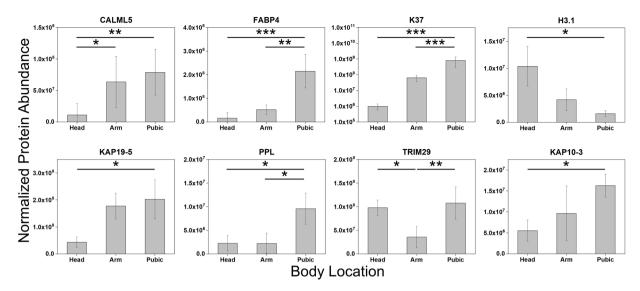


Figure 2. Average abundances for a subset of differentially expressed hair proteins at different body locations (two-way ANOVA and Tukey HSD; n=36). Error bars represent standard deviation from 4 replicate measurements of each of three individuals. Black lines represent statistically significant comparisons and significance levels are represented as: $p \le 0.05$ (*), $p \le 0.01$ (**), and $p \le 0.001$ (***).

is shown in Fig. 2 (two-way ANOVA and Tukey HSD). Further, many differentially expressed proteins show higher expression in pubic hair and are least abundant in arm hair, suggesting that pubic hair not only comprises a complex set of proteins, but also that proteins are more abundant in pubic hair compared to head and arm hair, even after accounting for mass differences. Not surprisingly, keratins and KAPs comprise only 27% of body location-specific differentially expressed proteins (i.e., 10 proteins), while intracellular proteins such as FABP4, MIF, and ATP5B make up the majority. As keratins and KAPs primarily contribute to the structural integrity of hair, which is highly conserved, it is unlikely that many hair structural proteins would exhibit differential expression at the various body locations. Many intracellular proteins are also least abundant in arm hair, although arm hair samples have notably high abundances of CALML5, GSDMA, and KAP19-5 compared to head hair samples. While the protein abundance profiles of head and arm hair samples are more similar compared to pubic hair, protein abundance variation in 37 markers enabled distinction of hair fibers from different body locations via principal components analysis (Supplementary Fig. S2). Differential protein expression captured with protein abundance confirms proteomic variation in hair from different body locations.

Effects of proteomic variation on GVP identification. Because protein abundances vary for a subset of hair proteins at different body locations and GVPs result from hair protein digests, it was considered that GVP identification may be affected by body location-specific differential protein expression. Therefore, it was imperative to examine the SNPs identified in each sample and determine whether differential protein expression affects GVP identification and subsequent SNP inference. Further comparison of identified SNPs in each sample was performed to observe whether some SNPs are only identified at specific body locations. Only SNP inferences consistent with an individual's genotype determined from exome sequencing were considered. SNPs with false positive responses are not robust candidates for a GVP panel and were removed; 65 SNPs remained for further analysis.

To observe any localization of SNPs, distributions of inferred SNPs from major and minor GVPs were compared across body locations. Of 65 SNPs, only exome-proteome consistent SNPs, in which the proteomic response corresponded with the exome response, i.e., true positive and true negative responses, across all 12 samples per body location for either major or minor GVPs, were retained (Fig. 3). Figure 3a,b illustrate the amount of overlap in consistent SNPs across samples from different body locations. From 11 and 14 consistent SNPs identified from major and minor GVPs, respectively, 5 and 8 SNPs are identified at all body locations, which comprise the majority (on average, 69%) of exome-proteome consistent SNPs. This observation suggests that reliable SNP identification in samples within a body location often extends to all samples. Only 11 SNPs in total are not identified at all body locations; there is one unreliably identified SNP that overlaps between major and minor GVPs.

The possibility that body location-specific SNP localization results from proteomic variation was further examined by comparing subsets of proteins. The subset of 37 proteins with body location-specific differential expression was compared with the proteins of 11 inconsistently identified SNPs (Fig. 3c). Any overlap in composition would indicate that differential expression of the protein affects downstream GVP identification and SNP inference within that protein. However, no overlap existed between differentially expressed proteins and proteins containing unreliably identified SNPs. With the exception of five proteins (APOD, CALML5, GSDMA, K37, KAP10-3), SNPs are not identified in body location-specific differentially expressed proteins. Despite significant positive correlations between the frequency of identifying SNPs from 3 of these proteins and protein abundance (Pearson product-moment correlation; $p \le 0.043$; Supplementary Fig. S3), identification of these SNPs remains

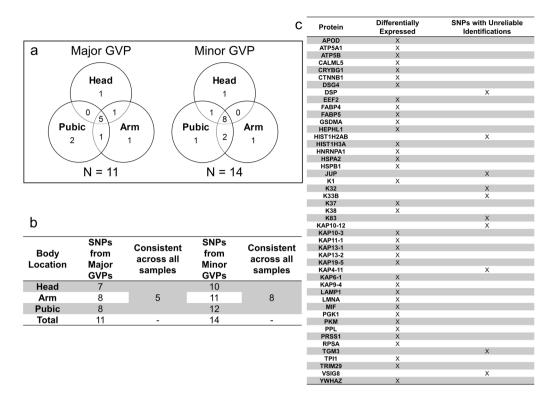


Figure 3. Comparison and distribution of exome-proteome consistent SNPs across different body locations. (a) Distribution of inferred consistent SNPs across the three body locations for major and minor GVPs, respectively. (b) Summary of the number of consistent SNPs at each body location. (c) Comparison of differentially expressed proteins to proteins of 11 SNPs with unreliable identifications at one or two body locations (i.e., not identified at all body locations). The majority of exome-proteome consistent SNPs identified at each body location are identified in all samples. Unreliably identified SNPs at either one or two body locations originate from a set of proteins that are not differentially expressed; there is no overlap between these sets of proteins. Therefore, SNPs are not body location-specific.

variable among sample replicates, regardless of body location. Further, no statistical positive correlation between SNP identification frequency and protein abundance was found for unreliably identified SNPs (Supplementary Fig. S3), demonstrating that body location-specific differential expression is not linked to SNP identification for all exome-proteome consistent SNPs. Therefore, while expression of APOD, GSDMA, and K37 may display some correlation with SNP identification, the vast majority (on average, 97%) of GVP identification is not affected by differential protein expression, especially if the peptides are consistently identified among sample replicates. SNP identification in hair specimens is not dependent on body location. GVP identification from protein digests of hair specimens is equally viable regardless of body location origin and all detected GVPs are candidates for a GVP panel.

GVP candidates for human identification panel. A series of criteria were established to evaluate GVP candidates for a robust panel. First, only GVPs that indicate exome-proteome consistent SNPs were considered. Furthermore, only consistent SNPs identified in all samples were selected, as these SNPs have the lowest false negative rates and their GVP counterparts have the highest chance of being detected. After accounting for overlap between major and minor GVPs, 12 SNPs remained for consideration. SNP identifiers, the two most abundant forms of the GVP, and their MS scan precursor ion abundances are reported in Table 1. See Supplementary Table S1 for a complete list of GVPs.

The second criterion used to evaluate GVP candidates in Table 1 is marker independence for random match probability (RMP) determination at the SNP level. To assess the performance of a robust panel for forensic identifications in a population, random match probabilities are calculated as the product of genotype frequencies for each SNP locus. However, genotype frequencies for correlated SNPs, i.e., SNPs in linkage disequilibrium ^{35,36}, may be biased in the population, which violates the assumption of marker independence for RMP calculations. To reduce the effect of possible disequilibria, a conservative one-SNP-per-gene rule was adopted; more sophisticated treatment of linkage disequilibrium will allow for inclusion of more GVPs, and thus, lower RMPs. For multiple SNPs from a gene, the SNP with the lowest minor allele frequency was selected. Finally, SNPs without Reference SNP IDs were also not considered further, as genotype frequencies are not known for these candidates. After applying these criteria, 8 SNPs remained for inclusion in a panel from 245 GVPs.

GVP profiles and identification performance. GVP profiles for each sample were established using 8 robust SNPs. Each GVP profile was established using the presence or absence of the major and minor GVPs at

Gene	SNP Identifier	Amino Acid Polymorphism	GVP Type	Peptide [†]	PTM	Average Abundance	Observation Frequency
FAM83H	rs9969600	Q/H	Minor	R.VNL H HVDFLR		1.10×10^{6}	2
KRT32	rs2071563	T/M	Major	R.ARLEGEINTYR	A1:Formylation	5.72×10^{7}	27
KRT32	rs2071563	T/M	Major	R.LEGEINTYR		1.71×10^{8}	28
KRT32	rs2071563	T/M	Minor	R.ARLEGEINMYR	M9:Oxidation (M)	4.79×10^{7}	10
KRT32	rs2071563	T/M	Minor	R.LEGEINMYR	M7:Oxidation (M)	1.69×10^{7}	7
KRT33A	rs148752041	D/H	Minor	R.HNAELENLIR	N2:Deamidation (NQ)	4.38×10^{7}	8
KRT33A	rs148752041	D/H	Minor	R.HNAELENLIRER		1.11×10^{8}	7
KRT33B	17:g.41366553 G > T *	L/M	Major	R.ILDELTLCRSDLEAQMESLKEELLSLKQ NHEQEVNTLR	C8:Carbamidomethylation;N29:Deamidation (NQ)	9.11×10^{7}	12
KRT33B	17:g.41366553 G > T *	L/M	Major	R.ILDELTLCRSDLEAQMESLKEELLSLK	C8:Carbamidomethylation	1.49×10^{7}	11
KRT33B	17:g.41366553 G > T *	L/M	Minor	R.ILDEMTLCR	C8:Carbamidomethylation	7.10×10^{7}	10
KRT33B	17:g.41366553 G > T *	L/M	Minor	R.RILDEMTLCR	C9:Carbamidomethylation	2.20×10^{7}	7
KRT81	rs2071588	G/R	Minor	R.GLTGGFGSHSVC R	C12:Carbamidomethylation	3.35×10^{8}	19
KRT81	rs2071588	G/R	Minor	L.TGGFGSHSVC R	C10:Carbamidomethylation	1.18×10^7	13
KRT83	rs2852464	I/M	Major	R.DLNMDCIVAEIK	C6:Carbamidomethylation	1.08×10^{8}	32
KRT83	rs2852464	I/M	Major	R.DLNMDCIVAEIKAQYDDIATR	K12:Carbamylation	1.83×10^{8}	23
KRT83	rs2852464	I/M	Minor	R.DLNMDC M VAEIK	C6:Carbamidomethylation	2.97×10^{7}	19
KRT83	rs2852464	I/M	Minor	R.DLNMDCMVAEIKAQYDDIATR	C6:Carbamidomethylation;M7:Oxidation (M)	4.29×10^{6}	16
KRTAP10-3	rs233252	C/Y	Minor	R.STYCVPIPSC	C4:Carbamidomethylation;C10:Carbamidomethylation	2.97×10^{6}	4
KRTAP10-3	rs233252	C/Y	Minor	R.STYCVPIPS	C4:Carbamidomethylation	1.65×10^{6}	2
KRTAP10-9	rs9980129	R/C	Minor	C.CAPTSSCQPSYCR	C1:Carbamidomethylation;C7:Carbamidomethylation; C12:Carbamidomethylation	6.99 × 10 ⁶	10
KRTAP4-11	rs760092771	S/C	Major	R.TTYCRPSCCVS S	C4:Carbamidomethylation;C8:Carbamidomethylation; C9:Carbamidomethylation	1.75×10^{8}	5
KRTAP4-11	rs760092771	S/C	Minor	R.TTYCRPSYSVSCC	C4:Carbamidomethylation;C12:Carbamidomethylation;C13:Carbamidomethylation	1.32×10^8	12
KRTAP4-11	rs760092771	S/C	Minor	R.TTYCRPSYSVSC	C4:Carbamidomethylation;C12:Carbamidomethylation	8.10×10^7	11
KRTAP4-11	rs763737606	C/S	Major	R.TTYCRPSCCVSS	C4:Carbamidomethylation;C8:Carbamidomethylation;C9:Carbamidomethylation	1.75×10^8	5
KRTAP4-11	rs763737606	C/S	Minor	R.TTYCRPSYSVSCC	C4:Carbamidomethylation;C12:Carbamidomethylation;C13:Carbamidomethylation	1.32×10^8	12
KRTAP4-11	rs763737606	C/S	Minor	R.TTYCRPSY S VSC	C4:Carbamidomethylation;C12:Carbamidomethylation	8.10×10^7	11
KRTAP4-11	rs774046661	C/Y	Major	R.TTYCRPSCCVSS	C4:Carbamidomethylation;C8:Carbamidomethylation;C9:Carbamidomethylation	1.75×10^{8}	5
KRTAP4-11	rs774046661	C/Y	Minor	R.TTYCRPSYSVSCC	C4:Carbamidomethylation;C12:Carbamidomethylation;C13:Carbamidomethylation	1.32×10^{8}	12
KRTAP4-11	rs774046661	C/Y	Minor	R.TTYCRPSYSVSC	C4:Carbamidomethylation;C12:Carbamidomethylation	8.10×10^{7}	11
VSIG8	rs62624468	V/I	Major	R.LGCPYVLDPEDYGPNGLDIEWMQVNSDPAHHR	C3:Carbamidomethylation;N15:Deamidation (NQ)	1.36×10^{7}	18
VSIG8	rs62624468	V/I	Major	R.LGCPYVLDPEDYGPNGLDIEWMQVNSDPAHHRE NVFLSYQDKR	C3:Carbamidomethylation;N15:Deamidation (NQ);M22:Oxidation (M)	6.88×10^{6}	14
VSIG8	rs62624468	V/I	Minor	R.LGCPYILDPEDYGPNGLDIEWMQVNSDPAHHR	C3:Carbamidomethylation;N15:Deamidation (NQ);M22:Oxidation (M)	4.92×10^{6}	5
VSIG8	rs62624468	V/I	Minor	R.LGCPYILDPEDYGPNGLDIEWMQVNSDPAHHRE NVFLSYQDKR	C3:Carbamidomethylation;N15:Deamidation (NQ);M22:Oxidation (M)	2.90×10^6	3

Table 1. SNP and GVP candidates for GVP panel. *No SNP identifier associated with SNP (HGVS notation used); Bold text denotes location of amino acid variant in genetically variant peptide; †Preceding amino acid in peptide sequence denoted by "X."

each SNP locus. Figure 4 displays a simplified version of each profile by using observed phenotype frequencies to represent the presence or absence of GVPs, as described in Materials and Methods. The full set of profiles that denotes the presence or absence of GVPs is found in Supplementary Fig. S4.

GVP profiles within an individual, irrespective of body location, are more similar compared to GVP profiles between individuals. Pairwise comparisons of GVP profiles allowed quantification of profile similarity, using the number of observed phenotype differences across 8 SNP loci, termed GVP profile differences. Differences were recorded if the compared responses did not match exactly, and then summed for each pairwise comparison, totaling 630 comparisons. Replicate comparisons, performed between hair specimens from the same individual and body location, yielded 1.17 ± 0.99 GVP profile differences, and within-individual comparisons, between hair samples from the same individual but different body locations, showed 1.06 ± 0.94 differences. As expected, between-individual comparisons exhibited the greatest number of GVP profile differences, with 4.92 ± 0.84 , 5.11 ± 0.92 , and 2.79 ± 0.71 differences, respectively, between Individuals 1-2, 2-3, and 1-3 (Fig. 5a). All observed profile differences approximate expected GVP profile differences (Fig. 5b). Greatest profile variation lies between individuals (Kruskal-Wallis test; $p=2.96 \times 10^{-108}$), demonstrating that despite some sample replicate and within-individual variation (e.g., body location), distinct GVP profiles are observed in samples from different individuals.

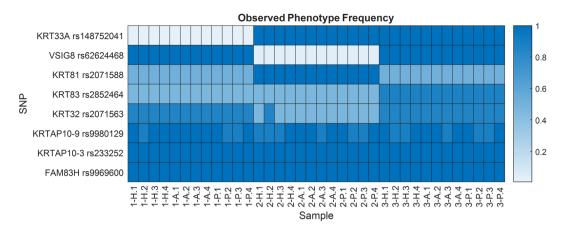


Figure 4. GVP profiles of 36 samples using observed phenotype frequency to represent the presence or absence of major and minor GVPs at 8 SNP loci. Profiles within an individual are similar, indicating consistent identification of SNPs with robust GVPs.

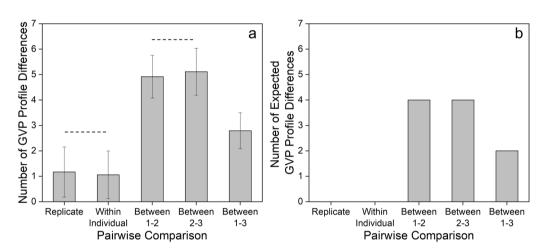


Figure 5. (a) Average number of GVP profile differences from different pairwise comparison categories compared to (b) expected number of GVP profile differences. Error bars represent the standard deviation. All but two comparisons, denoted by dotted line, are statistically significant (Kruskal-Wallis and Dunn tests; $n=630; p \leq 3.80 \times 10^{-6}$). The numbers of observed profile differences approximate expected GVP profile differences. Between Individual profile differences are statistically greater than Replicate and Within Individual profile differences.

Furthermore, RMPs, derived as products of observed phenotype frequencies from GVP profiles of each sample, align with the individual (Fig. 6). Experimental RMPs range between 1 in 3 and 1 in 870, within an order of magnitude of expected RMPs for each individual. Most importantly, GVP profiles of samples belonging to the same individual enable distinction of the individual to the same extent, regardless of body location, demonstrating that with a robust panel of inferred SNPs from GVPs, the probative value of one-inch head, arm, and pubic hair samples is equivalent within an individual.

Discussion

This study sought to examine body location-specific proteomic variation and its effects on SNP identification, and quantify the extent to which individuals are distinguished using single one-inch hairs from different body locations.

Many more proteins exhibited higher abundances in pubic hair compared to head and arm hair, highlighting the rich protein profile that can be obtained with pubic hair. Laatsch *et al.* reported differential expression levels for 12 proteins between head and pubic hair using spectral counts²⁴. Of these, 6 proteins had statistically different abundances between head and pubic hair samples in the current study, indicating agreement between studies using different protein quantification methods. In particular, K37 was found to be differentially expressed between head and pubic hair and is also statistically more abundant in pubic hair compared to arm hair. The presence of this protein in head hair, albeit at low levels, agrees with findings by previous studies^{24,26,28}. We observed unique peptides with a considerable abundance from this protein in only 25% of head hairs, while K37 was consistently expressed in arm and pubic hair. Because the hair samples prior to segmentation were of similar length

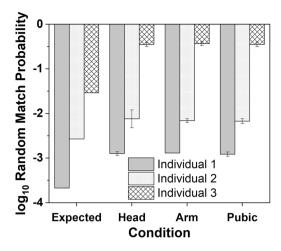


Figure 6. Experimentally observed random match probabilities ($m \pm 95\%$ CI) compared to expected RMP values for each individual. Expected RMPs are theoretically-derived values based on the detection of all GVPs consistent with an individual's genotype for the same 8 SNPs. RMP values of different body location samples from the same individual are not different; the extent to which individuals are distinguished from one another is not affected by hair origin. Observed RMP values from a robust set of SNPs approximate expected values within an order of magnitude.

(1–2 inches), absence of K37 in many head hairs cannot be attributed to degradation with age of hair, i.e., time since biosynthesis. Instead, K37 expression is linked to hair follicle keratinocyte differentiation differences at the various body locations; the protein is known to be expressed in the medulla of sexual hairs, including pubic hair, matured from unmedullated vellus hairs³⁷.

Hair structural proteins make up only 27% of differentially expressed proteins (i.e., 10 proteins), in agreement with Laatsch et al. 24. Variation in KAP abundance is more prevalent compared to keratins or other peripheral structural proteins. For example, KAP19-5 is highly abundant in arm and pubic hair compared to head hair, and pubic hair is significantly enriched with many other KAPs, including KAP11-1 and KAP10-3 (two-way ANOVA and Tukey HSD; Supplementary Fig. S5). Differential expression of KAPs may be linked to the structural conformation of intermediate filaments and affect physicochemical properties (e.g., rigidity, tensile strength, thickness) of hair fibers, and can serve as useful markers to differentiate hair fibers from different body locations (Supplementary Fig. S2). Distinction of head, arm, and pubic hair is further enhanced by differential expression of intracellular proteins, which dominates the hair proteome8. Protein expression is much more similar between head and arm hair; notably, histone H3.1 exhibits higher expression in head and arm hair compared to pubic hair and is a key protein in differentiating these hair types (Supplementary Fig. S2). Found in the nucleosome, the protein is an integral component in chromatin structure and is linked to DNA synthesis and repair³⁸, but also exhibits antimicrobial activity²⁸. As histone H3 is localized to the cortex of hair shaft³⁹, presence of variant H3.1 in head hair in high abundance is reasonable. Differentiation of hair fibers by body location based upon differential protein expression may be a valuable tool for screening in forensic investigations, but examination for any downstream effects on GVP and SNP identification is critically important for our forensic applications.

Effects of body location-specific proteomic variation on SNP identification were assessed by comparing protein subsets. Of the 37 differentially expressed proteins, 14% contain SNPs with variable exome-proteome consistency between sample replicates (Supplementary Fig. S3). These 8 SNPs comprise a minority, 12.3%, of all identified SNPs excluding false positives, and four SNPs show statistically significant positive correlations between protein abundance and frequency of response, including SNPs from K37. However, these SNPs are often not consistently identified among sample replicates, and in many cases, are not identified regardless of protein abundance, indicating that their absence is more likely attributed to a combination of the chemical characteristics of the GVP that suppresses their ionization and precursor ion competition for data-dependent MS/MS. Given the complexity of hair and that only the 10 most abundant ions per MS scan are fragmented, it is possible that the GVP is not sufficiently abundant to be selected for MS/MS in some samples. A key advantage of using data-dependent mass spectrometry is its breadth of proteome coverage, crucial in GVP discovery, but slight run-to-run differences in chromatographic separation may result in some irreproducibility in peptide selection for fragmentation of the chemical characteristics of the GVP panels will eliminate unreliable SNP identification.

All exome-proteome consistent SNPs derive from hair structural proteins that do not exhibit body location-specific differential protein expression (Fig. 3), and no significant positive correlation exists between these SNPs and proteins (Supplementary Fig. S3). While SNPs from APOD, GSDMA, and K37 show some correlation with body location-specific protein abundance differences, there is no evidence to support universal SNP localization from body location-specific proteomic variation. For 97% of markers, SNP identification is independent of protein expression differences in head, arm, and pubic hair.

GVP candidates shown in Table 1 represent the peptides observed with the highest frequency in this set of single hairs. Notably, most GVPs are tryptic; however, many peptides from KAPs are non-tryptic. Instead, some

of these peptides are cleaved at the C-terminal sides of cysteines or serines. Upon examination of their protein sequences, it is not surprising that non-tryptic GVPs are identified in KAPs, as there are few sites for trypsin to cleave among repeating chains of cysteine and serine near arginine-proline units. Because few tryptic sites exist, the frequency of identifying GVPs from KAPs is low compared to keratins¹⁹. Additionally, GVPs identified in structural proteins also contained missed cleavages, where the peptide contains one or more trypsin cleavage sites within its sequence. The presence of these peptides points to incomplete protein digestion by trypsin. Lower digestion efficiency in keratins is likely linked to their natural existence as coiled-coil structures through hydrophobic interactions¹. Hair keratins are stabilized as coiled-coil heterodimers with surfactant⁴²; removal of surfactant sodium dodecanoate during sample preparation likely resulted in the unfolding of hydrophobic proteins in a thermodynamically unfavorable environment, and subsequent protein aggregation⁴². Trypsin may have limited access to buried hydrophobic regions in keratins, thus reducing its digestion efficiency. However, a less efficient digestion can be beneficial when identifying unique peptides among these highly paralogous proteins^{4,5}.

GVP markers from 8 robust SNPs yield highly consistent GVP profiles within an individual (Fig. 4 and Supplementary Fig. S4). Not surprisingly, the majority of GVPs derive from keratins and KAPs, though markers from keratins provide more discriminative power with respect to phenotype frequencies for RMP calculations. Greater discriminative power of GVPs from keratins arises from consistent identification of minor GVPs, as opposed to the more variable identification of peptide markers in KAPs for reasons discussed above. Sporadic GVP identification also contributes to sample replicate and within-individual variation, although, as expected, interindividual variation dominates GVP profile differences, ranging between 3 and 5 times that of intraindividual variation. The highly similar intraindividual GVP profiles and RMPs demonstrate not only robustness in identifying these particular GVP markers, but body location invariance with this protein-based strategy.

We determined equivalent probative value of head, arm, and pubic hair for protein-based human identification using genetically variant peptide markers in this study. Body location-specific proteomic variation was characterized, with GVP identification and SNP inference invariant across hair from different body locations. Further, a set of robust SNPs inferred from exome-proteome consistent GVPs yields similar differentiative potential of individuals from hair specimens irrespective of body location. While future development of targeted mass spectrometry methods for GVP identification will eliminate intraindividual variation, current methods successfully demonstrated body location-specific GVP invariance in single one-inch hairs, defined criteria for marker selection, and identified robust GVP markers for protein-based human identification.

Materials and Methods

Hair sample preparation for mass spectrometry. Head, arm, and pubic hair specimens from three subjects (ages 25, 31, and 35) were collected to profile the protein variation in non-chemically treated hair from different body locations, under approval by the Institutional Review Board at Lawrence Livermore National Laboratory (IRB ID# 15-008) and in accordance with the Common Rule. Written informed consent for specimen collection and analysis was obtained prior to collection. Samples were stored in the dark at room temperature (RT). A one-inch (25 mm) single hair was segmented from a hair sample, and each was further segmented into four pieces of equal length (\sim 6 mm) for full immersion into the denaturation solution for protein extraction. To account for biological variation within individuals, different 25-mm single hair specimens from each body location were prepared as n=4 for each individual. The same protocol was followed to prepare the first three sets of replicates for proteomics-only analysis; a fourth set of replicates was prepared with a slightly modified protocol for protein/DNA co-extraction.

Aliquots of $100\mu L$ of an aqueous denaturation solution ($50\,mM$ dithiothreitol (DTT), $50\,mM$ ammonium bicarbonate (ABC), and $20\,mg/mL$ sodium dodecanoate (SDD)) were added to each single-inch hair specimen contained in a microcentrifuge tube. The tubes were sealed, placed in a $70\,^{\circ}C$ water bath, and ultrasonicated at a frequency of $37\,MHz$ and 100% power (Elma, Singen, Germany) until each hair sample was entirely solubilized (on average, 2h). $10\,\mu L$ of $1\,M$ iodoacetamide was added to each sample and the extracts were incubated in the dark at RT for $45\,min$.

To remove SDD (an ionization-suppressing compound in LC-MS/MS analysis) from the extracts, liquid-liquid extraction was performed using $100\,\mu\text{L}$ of 0.75% (v/v) trifluoroacetic acid in ethyl acetate. The upper organic layer was removed after phase separation and each extract was re-adjusted to pH = 8 with $10\,\mu\text{L}$ of $1\,\text{M}$ ABC. Protein concentration was performed using spin filter concentrators with a lock volume of $20\,\mu\text{L}$ (PES, $10\,\text{kDa}$ MWCO; Thermo Fisher Scientific Inc., Waltham, MA). Samples were centrifuged at $3,000\times g$ for $15\,\text{min}$ at RT, and $60\,\mu\text{L}$ of buffer solution ($50\,\text{mM}$ DTT and $50\,\text{mM}$ ABC) were added to wash the retentates, followed by a wash with $30\,\mu\text{L}$ of buffer after centrifugation for $15\,\text{min}$. Finally, spin filter retentates were centrifuged for $30\,\text{min}$ and reconstituted to $50\,\mu\text{L}$ with buffer solution ($50\,\text{mM}$ DTT, $50\,\text{mM}$ ABC, and $0.1\,\text{mg/mL}$ ProteaseMAXTM Surfactant (Promega, Madison, WI)) prior to overnight trypsin digestion (TPCK-treated, sequencing grade) for at least $18\,\text{h}$ accompanied by magnetic stirring at RT.

Protein digests were filtered for particulates using centrifugal filter tubes (PVDF, $0.1\,\mu m$; Millipore Sigma, Burlington, MA) with centrifugation at $9,000\times g$ for $10\,m$ at RT. Filtered digests for the first three sets of replicates were then analyzed by LC-MS/MS. For the fourth set of replicates, a protein/DNA co-extraction procedure was performed. $200\,\mu L$ of ethanol was added to each filtrate, and the mixture was transferred to a DNA-binding column from the QIAamp DNA Micro Kit (Qiagen, Hilden, Germany) and fractionated via centrifugation into a protein fraction (flow-through) and a DNA fraction (retentate). Protein digest eluate collected after centrifugation at $6,000\times g$ for $1\,m$ m at RT was evaporated to dryness under vacuum and reconstituted in $50\,\mu L$ of $50\,m$ MABC, $50\,m$ DTT, and $0.1\,m$ g/mL ProteaseMAXTM Surfactant. The reconstituted protein digest was filtered as above and analyzed via LC-MS/MS. Analysis and results from the DNA fraction will be described in a future publication.

Liquid chromatography-tandem mass spectrometry analysis. Protein digests were analyzed on an EASY-nLC 1200 system coupled to a Q Exactive Plus Orbitrap mass spectrometer (Thermo Fisher Scientific

Inc., Waltham, MA). $1\,\mu\text{L}$ injections were loaded onto an AcclaimTM PepMapTM 100 C18 trap ($75\,\mu\text{m} \times 20\,\text{mm}$, $3\,\mu\text{m}$ particle size), washed, and separated on an Easy-SprayTM C18 analytical column ($50\,\mu\text{m} \times 150\,\text{mm}$, $2\,\mu\text{m}$ particle size). Separations were performed at a flow rate of $300\,\text{nL/min}$ using mobile phases A (0.1% formic acid in water) and B (0.1% formic acid in 90% acetonitrile/10% water) over a 107-min gradient: 2 to 3% B in 1 min, 3 to 11% B in 75 min, 11 to 39% B in 15 min, ramped to 100% B in 1 min, and held at 100% B for 15 min. Positive mode nano-electrospray ionization was achieved at a voltage of $1.9\,\text{kV}$. Full MS scans were acquired at a resolution of 70,000, with a maximum ion accumulation time of 30 ms, and a scan range between m/z 380 and 1800. Data-dependent MS/MS scans were triggered for the 10 most abundant ions at an intensity threshold of 3.3×10^4 and acquired at a resolution of 17,500, with a maximum ion accumulation time of 60 ms, dynamic exclusion of 24 s, and an isolation window of 2 Da. HCD fragmentation was performed at a collision energy setting of 27. Singly-charged species and ions with unassigned charge states were excluded from MS/MS.

Protein and peptide identification. Mass spectral data were imported into PEAKS Studio 8.5 (Bioinformatics Solutions Inc., Waterloo, ON, Canada) for peptide identification via de novo sequencing and subsequent database searching. Precursor ion mass tolerance was selected as ± 20 ppm, while a mass error of 0.05 Da was allowed for fragment ions. A list of 313 potential post-translational modifications, which includes cysteine carbamidomethylation, methionine oxidation, and asparagine and glutamine deamidation, was allowed as variable modifications for peptide identification. The maximum number of PTMs allowed per peptide was three, and 3 tryptic missed cleavages on either end of the peptide were permitted. All de novo-sequenced peptides with a confidence score $(-10l \,\mathrm{gP})$ greater than 15% were matched to protein sequences in a reference database. To capture non-mutated proteins in the samples, the UniProtKB SwissProt Human protein database (downloaded September 27, 2017) was used for protein inference from identified peptides⁴³. A second protein and peptide identification using the same raw mass spectral files was performed in PEAKS, where de novo-sequenced peptides were searched against individualized protein databases created from exome sequence data (Supplementary Methods). The second PEAKS analysis enabled a focused search for proteins with expected mutations to identify GVPs in each sample. Each individualized protein database contains protein sequences from a list of 691 common gene products found in hair with the appropriate mutations expected in an individual based on their exome sequence. GVPs identified from each hair specimen were matched to mutated protein sequences in individualized protein databases in the second peptide identification analysis.

Identified proteins and peptides were further filtered with a 1% false-discovery rate threshold for peptide-spectrum matches and then exported from PEAKS. An in-house Python-based script was applied to the output files to merge results from the two PEAKS analyses and generate a non-redundant protein profile for each sample. Protein profile metrics include the number of proteins, unique peptide sequences, amino acids, and SNPs identified from both major and minor GVPs.

GVP profile generation – observed phenotype frequencies. Each GVP profile was established using the presence or absence of the major and minor GVPs at each SNP locus. Observed phenotype frequencies were used to represent the presence or absence of GVPs. Conventionally, SNPs are associated with population genotype frequencies, obtained from the Genome Aggregation Database (gnomAD)⁴⁴. However, to account for uncertainty in establishing a genotype with proteomic responses from incomplete GVP detection, observed phenotype frequencies were used as sums of genotype frequencies (i.e., sum of either major homozygote or minor homozygote genotype frequency with heterozygote genotype frequency) when only either a major or minor GVP was detected. Total population genotype frequencies from gnomAD were used. For example, detection of only the minor GVP for a SNP resulted in an observed phenotype frequency as the sum of the heterozygote and minor homozygote frequencies. Observed phenotype frequency was not reported for absent GVPs (i.e., one true negative and one false negative response) as the SNP was not considered in that sample.

Statistical analysis. All statistical comparisons were performed in R (x64 version 3.4.4). Significance was established at $\alpha = 0.05$. Two-way ANOVAs were performed using the *aov* function after fitting a linear model in the *stats v3.5.0* package. Tukey HSD post-hoc tests, two sample t-tests, and tests for association of Pearson product-moment correlations were performed using the same package. Equal variances were not assumed for t-tests. For non-parametric Kruskal-Wallis and Dunn post-hoc tests, the *agricolae v.1.2–8* package and *dunn. test v.1.1.0* package, respectively, were used, and a Bonferroni correction was applied to adjust p-values. Principal components analysis (PCA) was performed in MATLAB (R2017a, MathWorks, Natick, MA). All plots were drawn in OriginPro 2018 (OriginLab Corp., Northampton, MA) except for MATLAB outputs; PCA plots were drawn in Microsoft Excel 2016 (Redmond, WA). All values are reported as m \pm s.d. unless otherwise specified.

Disclaimer. The Lawrence Livermore National Laboratory, Office of Scientific and Technical Information, Information Management (IM) number is: LLNL-JRNL-757511. This document was prepared as an account of work sponsored by an agency of the United States government. Neither the United States government nor Lawrence Livermore National Security, LLC, nor any of their employees makes any warranty, expressed or implied, or assumes any legal liability or responsibility for the accuracy, completeness, or usefulness of any information, apparatus, product, or process disclosed, or represents that its use would not infringe privately owned rights. Reference herein to any specific commercial product, process, or service by trade name, trademark, manufacturer, or otherwise does not necessarily constitute or imply its endorsement, recommendation, or favoring by the United States government or Lawrence Livermore National Security, LLC. The views and opinions of authors expressed herein do not necessarily state or reflect those of the United States government or Lawrence Livermore National Security, LLC, and shall not be used for advertising or product endorsement purposes.

Data Availability

The mass spectrometry proteomics data generated during the current study have been deposited to the Proteom-eXchange Consortium (http://proteomecentral.proteomexchange.org) via PRIDE with identifier PXD010982.

References

- 1. Zhang, Y., Alsop, R. J., Soomro, A., Yang, F.-C. & Rheinstädter, M. C. Effect of shampoo, conditioner and permanent waving on the molecular structure of human hair. *Peer*] 3, e1296, https://doi.org/10.7717/peerj.1296 (2015).
- Schweizer, J., Langbein, L., Rogers, M. A. & Winter, H. Hair follicle-specific keratins and their diseases. Exp Cell Res 313, 2010–2020, https://doi.org/10.1016/j.yexcr.2007.02.032 (2007).
- 3. Wolfram, L. J. Human hair: A unique physicochemical composite. J Am Acad Dermatol 48, \$106–\$114, https://doi.org/10.1067/mjd.2003.276 (2003).
- Langbein, L. et al. The catalog of human hair keratins: I. Expression of the nine type I members in the hair folicle. J Biol Chem 274, 19874–19884, https://doi.org/10.1074/jbc.274.28.19874 (1999).
- 5. Langbein, L., Rogers, M. A., Winter, H., Praetzel, S. & Schweizer, J. The catalog of human hair keratins: II. Expression of the six type II members in the hair folicle and the combined catalog of human type I and II keratins. *J Biol Chem* **276**, 35123–35132, https://doi.org/10.1074/jbc.M103305200 (2001).
- Rouse, J. G. & Van Dyke, M. E. A review of keratin-based biomaterials for biomedical applications. *Materials* 3, https://doi. org/10.3390/ma3020999 (2010).
- Shimomura, Y. & Ito, M. Human hair keratin-associated proteins. J Investig Dermatol Symp Proc 10, 230–233, https://doi. org/10.1111/j.1087-0024.2005.10112.x (2005).
- 8. Lee, Y. J., Rice, R. H. & Lee, Y. M. Proteome analysis of human hair shaft: From protein identification to posttranslational modification. *Mol Cell Proteomics* 5, 789–800, https://doi.org/10.1074/mcp.M500278-MCP200 (2006).
- Szabo, S. et al. In situ labeling of DNA reveals interindividual variation in nuclear DNA breakdown in hair and may be useful to
 predict success of forensic genotyping of hair. Int J Legal Med. 126, 63–70, https://doi.org/10.1007/s00414-011-0566-5 (2012).
- McNevin, D., Wilson-Wilde, L., Robertson, J., Kyd, J. & Lennard, C. Short tandem repeat (STR) genotyping of keratinised hair Part
 An optimised genomic DNA extraction procedure reveals donor dependence of STR profiles. Forensic Sci Int 153, 247–259, https://doi.org/10.1016/j.forsciint.2005.05.005 (2005).
- 11. Butler, J. M., Buel, E., Crivellente, F. & McCord, B. R. Forensic DNA typing by capillary electrophoresis using the ABI Prism 310 and 3100 genetic analyzers for STR analysis. *Electrophoresis* 25, 1397–1412, https://doi.org/10.1002/elps.200305822 (2004).
- 12. Butler, J. M. & Hill, C. R. Biology and genetics of new autosomal STR loci useful for forensic DNA. *analysis. Forensic Science Review* 24, 15–26 (2012).
- 13. Kidd, K. K. et al. Developing a SNP panel for forensic identification of individuals. Forensic Sci Int 164, 20–32, https://doi.org/10.1016/j.forsciint.2005.11.017 (2006).
- Pakstis, A. J. et al. SNPs for a universal individual identification panel. Hum. Genet. 127, 315–324, https://doi.org/10.1007/s00439-009-0771-1 (2010).
- 15. Kim, J.-J., Han, B.-G., Lee, H.-I., Yoo, H.-W. & Lee, J.-K. Development of SNP-based human identification system. *Int J Legal Med.* **124**, 125–131, https://doi.org/10.1007/s00414-009-0389-9 (2010).
- Greaves, M. P. & Wilson, M. J. The degradation of nucleic acids and montmorillonite-nucleic-acid complexes by soil microorganisms. Soil Biol and Biochem. 2, 257–268, https://doi.org/10.1016/0038-0717(70)90032-5 (1970).
- Ogram, A. V., Mathot, M. L., Harsh, J. B., Boyle, J. & Pettigrew, C. A. Effects of DNA polymer length on its adsorption to soils. Appl Environ Microbiol 60, 393–396 (1994).
- 18. Wadsworth, C. & Buckley, M. Proteome degradation in fossils: investigating the longevity of protein survival in ancient bone. *Rapid Commun Mass Spectrom* 28, 605–615, https://doi.org/10.1002/rcm.6821 (2014).
- 19. Solazzo, C. et al. Characterisation of novel α-keratin peptide markers for species identification in keratinous tissues using mass spectrometry. Rapid Commun Mass Spectrom 27, 2685–2698, https://doi.org/10.1002/rcm.6730 (2013).
- 20. Legg, K. M., Powell, R., Reisdorph, N., Reisdorph, R. & Danielson, P. B. Discovery of highly specific protein markers for the identification of biological stains. *Electrophoresis* 35, 3069–3078, https://doi.org/10.1002/elps.201400125 (2014).
- Legg, K. M., Powell, R., Reisdorph, N., Reisdorph, R. & Danielson, P. B. Verification of protein biomarker specificity for the identification of biological stains by quadrupole time-of-flight mass spectrometry. *Electrophoresis* 38, 833–845, https://doi. org/10.1002/elps.201600352 (2016).
- 22. Rice, R. H., Wong, V. J., Price, V. H., Hohl, D. & Pinkerton, K. E. Cuticle cell defects in lamellar ichthyosis hair and anomalous hair shaft syndromes visualized after detergent extraction. *Anat Rec* 246, 433-441, https://doi.org/10.1002/(SICI)1097-0185(199612)246:4<433::AID-AR2>3.0.CO;2-W (1996).
- 23. Rice, R. H. Proteomic analysis of hair shaft and nail plate. J Cosmet Sci 62, 229-236 (2011).
- Laatsch, C. N. et al. Human hair shaft proteomic profiling: individual differences, site specificity and cuticle analysis. Peer 2, e506, https://doi.org/10.7717/peerj.506 (2014).
- 25. Wu, P.-W. et al. Proteomic analysis of hair shafts from monozygotic twins: Expression profiles and genetically variant peptides. Proteomics 17, 1600462, https://doi.org/10.1002/pmic.201600462 (2017).
- Parker, G. J. et al. Demonstration of protein-based human identification using the hair shaft proteome. PLoS ONE 11, e0160653, https://doi.org/10.1371/journal.pone.0160653 (2016).
- 27. Mason, K. E., Paul, P. H., Chu, F., Anex, D. S. & Hart, B. R. Development of a Protein-based Human Identification Capability from a Single Hair. *Journal of Forensic Sciences* **0**, https://doi.org/10.1111/1556-4029.13995 (2019).
- Aday, S. S. et al. Studies on the proteome of human hair Identification of histones and deamidated keratins. Sci Rep 8, 1599, https://doi.org/10.1038/s41598-018-20041-9 (2018).
- 29. Wang, W. et al. Quantification of proteins and metabolites by mass spectrometry without isotopic labeling or spiked standards. Anal. Chem. 75, 4818–4826 (2003).
- 30. Liu, H., Sadygov, R. G. & Yates, J. R. A model for random sampling and estimation of relative protein abundance in shotgun proteomics. *Anal. Chem.* **76**, 4193–4201 (2004).
- 31. Old, W. M. *et al.* Comparison of label-free methods for quantifying human proteins by shotgun proteomics. *Mol Cell Proteomics* 4, 1487–1502 (2005).
- 32. Chelius, D. & Bondarenko, P. V. Quantitative profiling of proteins in complex mixtures using liquid chromatography and mass spectrometry. *J Proteome Res.* 1, 317–323 (2002).
- 33. Griffin, N. M. et al. Label-free, normalized quantification of complex mass spectrometry data for proteomic analysis. *Nature Biotechnol* 28, 83, https://doi.org/10.1038/nbt.1592 (2009).
- 34. Zhang, Y., Wen, Z., Washburn, M. P. & Florens, L. Effect of dynamic exclusion duration on spectral count based quantitative proteomics. *Anal. Chem.* 81, 6317–6326, https://doi.org/10.1021/ac9004887 (2009).
- 35. Reich, D. E. et al. Linkage disequilibrium in the human genome. Nature 411, 199, https://doi.org/10.1038/35075590 (2001).
- 36. Stephens, J. C. et al. Haplotype variation and linkage disequilibrium in 313 human genes. Science 293, 489 (2001).
- 37. Jave-suarez, L. F. et al. Androgen regulation of the human hair follicle: The type I hair keratin hHa7 is a direct target gene in trichocytes. J Investig Dermatol 122, 555–564, https://doi.org/10.1111/j.0022-202X.2004.22336.x (2004).

- 38. Tagami, H., Ray-Gallet, D., Almouzni, G. & Nakatani, Y. Histone H3.1 and H3.3 complexes mediate nucleosome assembly pathways dependent or independent of DNA synthesis. *Cell* 116, 51–61, https://doi.org/10.1016/S0092-8674(03)01064-X (2004).
- 39. Subbaiah, R. S. et al. Identification of antibacterial components in human hair shafts. Acta Derm Venereol 98, 708–710, https://doi.org/10.2340/00015555-2952 (2018).
- 40. Tabb, D. L. et al. Repeatability and reproducibility in proteomic identifications by liquid chromatography—tandem mass spectrometry. J Proteome Res. 9, 761–776, https://doi.org/10.1021/pr9006365 (2010).
- Bateman, N. W. et al. Maximizing peptide identification events in proteomic workflows utilizing data-dependent acquisition. Mol Cell Proteomics. https://doi.org/10.1074/mcp.M112.026500 (2013).
- 42. Schrooyen, P. M. M., Dijkstra, P. J., Oberthür, R. C., Bantjes, A. & Feijen, J. Stabilization of solutions of feather keratins by sodium dodecyl sulfate. *J Colloid Interface Sci* 240, 30–39, https://doi.org/10.1006/jcis.2001.7673 (2001).
- 43. Boutet, E., Liberherr, D., Tognolli, M., Schneider, M. & Bairoch, A. UniProtKB/Swiss-Prot. Methods Mol Biol 406, 89-112 (2007).
- 44. Lek, M. et al. Analysis of protein-coding genetic variation in 60,706 humans. Nature 536, 285, https://doi.org/10.1038/nature19057 (2016)

Acknowledgements

The authors thank Dr. Phillip H. Paul for providing annotated variant lists and individualized protein databases. This work was financially supported by the Department of Energy, Lawrence Livermore National Laboratory, Laboratory Directed Research and Development award (16-SI-002) and the Livermore Graduate Scholars Program. This work was performed under the auspices of the U.S. Department of Energy by Lawrence Livermore National Laboratory under Contract DE-AC52-07NA27344.

Author Contributions

All authors contributed to experiment design, F.C., K.E.M., and D.S.A. performed the experiment, B.R.H. provided resources and materials, and all authors contributed to the writing of the manuscript.

Additional Information

Supplementary information accompanies this paper at https://doi.org/10.1038/s41598-019-44007-7.

Competing Interests: The authors declare no competing interests.

Publisher's note: Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

Open Access This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons license, and indicate if changes were made. The images or other third party material in this article are included in the article's Creative Commons license, unless indicated otherwise in a credit line to the material. If material is not included in the article's Creative Commons license and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this license, visit https://creativecommons.org/licenses/by/4.0/.

© The Author(s) 2019