Mass-encoded synthetic biomarkers for multiplexed urinary monitoring of disease

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Biomarkers are becoming increasingly important in the clinical management of complex diseases, yet our ability to discover new biomarkers remains limited by our dependence on endogenous molecules. Here we describe the development of exogenously administered 'synthetic biomarkers' composed of mass-encoded peptides conjugated to nanoparticles that leverage intrinsic features of human disease and physiology for noninvasive urinary monitoring. These protease-sensitive agents perform three functions: in vivo: they target sites of disease, sample dysregulated protease activities and emit mass-encoded reporters into host urine for multiplexed detection by mass spectrometry. Using mouse models of liver fibrosis and cancer, we show that these agents can noninvasively monitor liver fibrosis and resolution without the need for invasive core biopsies and substantially improve early detection of cancer compared with current clinically used blood biomarkers. This approach of engineering synthetic biomarkers for multiplexed urinary monitoring should be broadly amenable to additional pathophysiological processes and point-of-care diagnostics.

RESULTS
Protease-sensitive nanoparticles for urinary monitoring

To develop a protease-sensing platform, we first identified peptide substrates of proteases implicated in liver fibrosis and cancer.

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We conjugated fluorescein-labeled derivatives of ~50 candidate peptide substrates to polyethylene glycol–coated, long-circulating iron oxide nanoworm nanoparticles (Supplementary Fig. 1a–c) and incubated them with recombinant proteases commonly overexpressed in disease (e.g., matrix metalloproteases (MMPs) and cathepsins) as well as blood-borne proteases to assess crossreactivity (e.g., FXa, tissue factor and thrombin). We determined the relative substrate activities for each protease-substrate combination by monitoring increases in sample fluorescence resulting from peptidolysis that allowed previously homoquenched fluorophores to emit freely in solution (Fig. 2a). We compiled the initial reaction velocities for comparative analysis in a heat map (Fig. 2b), from which we selected ten peptide substrates (S1–S10; Table 1) with broad protease susceptibility as our peptide-nanoworm library.

To establish the potential to probe disease microenvironments remotely from urine, we next investigated the in vivo behavior of each system component (i.e., peptide and nanoworm). We selected a xenobiotic mouse model of liver fibrosis in which FVB/NJ mice fed with 3,5-diethoxycarbonyl-1,4-dihydrocollidine (DDC) develop progressive liver disease as a result of chronic bile duct injury25, leading to liver fibrosis and upregulation of local MMPs26. Despite the multiplexing advantages of mass encoding, one challenge for each protease-substrate combination by promiscuous proteases and truncated by extraneous cleavage. To design an extensible encoding strategy for our library of protease substrates, we adapted principles of isobaric mass encoding to produce a family of mass reporters and promote renal filtration after substrate cleavage and release from nanoworms. We further modified these tandem peptides with internal photolabile residues to enable the recovery of Glu-fib peptides by photolysis from complex urinary cleavage fragments after in vivo proteolysis. To test this construct, we synthesized a model photolabeled tandem peptide (compound 1; Fig. 3a). Consistent with previously published reports on nitrophenyl groups, exposure of compound 1 (triphly charged, 881.7 mass/charge ratio (m/z); Fig. 3b) to UV light triggered peptide cleavage, resulting in the appearance of doubly charged, acetamide-terminated Glu-fib (785.4 m/z, Fig. 3b).

To validate this approach, we analyzed an equimolar 10-plex iCORE library of urinary biomarkers. We found that substrates in complex proteolytic environments can be cleaved at multiple sites by promiscuous proteases and truncated by exoproteases to produce diverse pools of poorly defined fragments that confound mass analysis. Here we created well-defined mass reporters to encode our substrate library. In light of the favorable renal clearance properties of Glu-fib, we appended δ isomer–rich derivatives of Glu-fib to the N terminus of each protease substrate to serve as protease-resistant mass reporters and promote renal filtration after substrate cleavage and release from nanoworms. We then developed a method that endogenously covalently modifies Glu-fib to the N terminus of each protease substrate to serve as protease-resistant mass reporters and promote renal filtration after substrate cleavage and release from nanoworms. We further modified these tandem peptides with internal photolabile residues to enable the recovery of Glu-fib peptides by photolysis from complex urinary cleavage fragments after in vivo proteolysis. To test this construct, we synthesized a model photolabeled tandem peptide (compound 1; Fig. 3a). Consistent with previously published reports on nitrophenyl groups, exposure of compound 1 (triphly charged, 881.7 mass/charge ratio (m/z); Fig. 3b) to UV light triggered peptide cleavage, resulting in the appearance of doubly charged, acetamide-terminated Glu-fib (785.4 m/z, Fig. 3b).

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library (R1–R10; Table 1) by liquid chromatography MS/MS (LC-MS/MS) and found the entire peptide library to be present initially as a single, unresolved peak (extracted ion chromatogram (XIC), 789.85 m/z; Fig. 3c,d) but then resolve after fragmentation as predicted into a ten-peak spectrum with no fragmentation bias (683.4–692.4 m/z; Fig. 3e and Supplementary Fig. 3). We removed confounding peak overlap from naturally occurring isotopes (e.g., 13C) by collecting iCORE peptides with a 1-μL window centered on the precursor ion (Supplementary Fig. 4a), which minimized the signal from naturally occurring isotopes to ~5% of the parent peak (Supplementary Fig. 4b). Consequently, in samples spiked with reporters at defined ratios (1:2:3:5:10:10:5:3:2:1), we found a linear correlation between peak intensity and stoichiometry in both unmodified and peak-subtracted analyses (n = 3 spiked samples, \( R^2 = 0.99 \) and \( R^2 = 0.99 \), respectively; Supplementary Fig. 5a–c). We peak adjusted all subsequent samples to reflect contributions from naturally occurring isotopes.

To test the ability of iCORE reporters for monitoring peptide cleavage, we extended protease substrates S1–S10 with iCORE mass tags R1–R10 using photosensitive amino acids and coupled them to nanoworms to produce synthetic biomarkers G1–G10 (Table 1). After treatment of an equimolar cocktail of G1–G10 with recombinant MMP9, we isolated peptide products by size filtration and exposed them to UV light to release reporters R1–R10 for MS/MS quantification. Collective substrate activities had distinct iCORE landscapes with individual \( y_6 \) peak intensities corresponding to the substrate preference for MMP9 (Fig. 3f). We applied this library to several additional proteases (Supplementary Fig. 6a) and found that their iCORE profiles were unique, as determined from Pearson’s correlation analysis (i.e., MMP2, MMP9, MMP12 and thrombin; Supplementary Fig. 6b), illustrating the ability of iCORE-encoded nanoworms to monitor many protease-substrate combinations simultaneously.

### Monitoring hepatic fibrogenesis and resolution

Liver fibrosis is a wound-healing response to chronic liver injury and results in the deposition of scar tissue that can lead to cirrhosis, liver failure and cancer17. The dynamics of extracellular matrix (e.g., collagen) accumulation are driven largely by activated hepatic stellate cells and matrix remodeling proteases such as MMPs and their inhibitors. The current gold standard for monitoring this process is a needle biopsy followed by histological analysis; however, this technique is invasive, confounded by high sampling heterogeneity, carries a finite risk of complications and cannot be performed as frequently as needed (e.g., for assessing antifibrotic therapies)14. Noninvasive assays, including ultrasound imaging, elastography and

![Figure 2](image)

**Figure 2** Urinary detection of in vivo protease activity with peptide-nanoworms. (a) Representative activation profiles of peptide-nanoworms after treatment with recombinant proteases. Specific protease-substrate combinations led to rapid increases in sample fluorescence. RFU, relative fluorescence units. (b) Heat map comparison of the cleavage velocities for different substrate-protease combinations grouped according to activity and specificity. NW, nanoworm; Fl-skGG, fluorescein-skGG. (c–e) Fluorescence in vivo images of DDC-treated and control mice after intravenous injection of VivoTag 680-labeled Glu-fib peptides (c), peptide-free nanoworms (d) or peptide-conjugated nanoworms (e). RES, reticuloendothelial system.

### Table 1 10-plex synthetic biomarker library

| Synthetic biomarker library a,b (G1–G10) | Substrate (S1–S10) | Isobaric mass code c,d (R1–R10) | \( y_6 \) reporter | \( y_6 + H^+ \) |
|---|---|---|---|---|
| e\(^3\)G\(^5\)G\(^6\)Vt\(^n\)neeG\(^f\)s\(^n\)Ar-X-K(FAM)G\(^G\)GP\(^G\)QW\(^G\)QGC-NW | PG\(^G\)W\(^Q\) | e\(^1\)G\(^2\)G\(^3\)Vt\(^n\)neeG\(^4\)f\(^s\)Ar | G\(^f\)s\(^n\)Ar | 683.4 |
| e\(^2\)G\(^5\)Vt\(^n\)neeG\(^f\)s\(^n\)Ar-X-K(FAM)G\(^G\)GP\(^G\)SR\(^G\)SGC-NW | LVR\(^R\)SGGC | e\(^1\)G\(^2\)G\(^3\)Vt\(^n\)neeG\(^4\)f\(^s\)Ar | G\(^f\)s\(^n\)Ar | 684.4 |
| e\(^1\)G\(^2\)G\(^5\)Vt\(^n\)neeG\(^f\)s\(^n\)Ar-X-K(FAM)G\(^G\)GP\(^G\)W\(^G\)GQC-NW | PV\(^G\)L\(^L\)IGGC | e\(^1\)G\(^2\)G\(^3\)Vt\(^n\)neeG\(^4\)f\(^s\)Ar | G\(^f\)s\(^n\)Ar | 685.4 |
| e\(^1\)G\(^3\)G\(^5\)Vt\(^n\)neeG\(^f\)s\(^n\)Ar\(^n\) Ar-X-K(FAM)G\(^G\)GP\(^G\)L\(^L\)R\(^L\)WS\(^W\)CN-CW | PG\(^L\)L\(^L\)RGG | e\(^1\)G\(^2\)G\(^3\)Vt\(^n\)neeG\(^4\)f\(^s\)Ar | G\(^f\)s\(^n\)Ar | 687.4 |
| e\(^3\)G\(^5\)Vt\(^n\)neeG\(^f\)s\(^n\)Ar-X-K(FAM)G\(^G\)GP\(^G\)LR\(^G\)GRC-NW | PL\(^L\)GR\(^R\)KG | e\(^1\)G\(^2\)G\(^3\)Vt\(^n\)neeG\(^4\)f\(^s\)Ar | G\(^f\)s\(^n\)Ar | 689.4 |
| e\(^2\)G\(^5\)Vt\(^n\)neeG\(^f\)s\(^n\)Ar\(^n\) Ar-X-K(FAM)G\(^G\)GP\(^G\)P\(^P\)S\(^S\)G\(^G\)C-CW | f(Pip)R\(^P\)SG\(^G\)G | e\(^1\)G\(^2\)G\(^3\)Vt\(^n\)neeG\(^4\)f\(^s\)Ar | G\(^f\)s\(^n\)Ar | 690.4 |
| e\(^1\)G\(^3\)G\(^5\)Vt\(^n\)neeG\(^f\)s\(^n\)Ar\(^n\) Ar-X-K(FAM)G\(^G\)GP\(^G\)RSS\(^G\)GC-NW | f\(^P\)R\(^P\)SG\(^G\)G | e\(^1\)G\(^2\)G\(^3\)Vt\(^n\)neeG\(^4\)f\(^s\)Ar | G\(^f\)s\(^n\)Ar | 691.4 |
| e\(^2\)G\(^5\)Vt\(^n\)neeG\(^f\)s\(^n\)Ar\(^n\) Ar-X-K(FAM)G\(^G\)G\(^G\)P\(^P\)RSS\(^G\)GC-NW | f\(^P\)R\(^P\)SG\(^G\)G | e\(^1\)G\(^2\)G\(^3\)Vt\(^n\)neeG\(^4\)f\(^s\)Ar | G\(^f\)s\(^n\)Ar | 692.4 |
| e\(^3\)G\(^5\)Vt\(^n\)neeG\(^f\)s\(^n\)Ar\(^n\) Ar-X-K(FAM)G\(^G\)GP\(^G\)P\(^P\)G\(^P\)RSS\(^G\)GC-NW | f\(^P\)R\(^P\)SG\(^G\)G | e\(^1\)G\(^2\)G\(^3\)Vt\(^n\)neeG\(^4\)f\(^s\)Ar | G\(^f\)s\(^n\)Ar | 692.4 |

List of synthetic biomarkers (G1–G10), protease substrates (S1–S10) and isobaric mass reporters (R1–R10) used in study.

*a* 3-amino-3-(2-nitrophenyl)propionic acid; FAM, carboxyfluorescein; Pip, pipecolic acid; NW, nanoworm. *b* Lowercase indicates the \( \delta \) isomer amino acid. *c* Photocleaved C terminus, CONH\(_2\). *d* Mass = 1,089.8 Da.
serum biomarkers, are limited by their low accuracies and limited prognostic utility\textsuperscript{35}. Thus, there remains an urgent need for non-invasive biomarkers to replace biopsy-based monitoring to facilitate the identification and validation of new anti-fibrotic agents and support clinical decision making\textsuperscript{36}. We sought to identify synthetic biomarkers with the capacity to monitor liver fibrosis and resolution and extended our DDC model to include both aspects of the disease.

We first evaluated the potential toxicity of nanoworms to determine whether serial monitoring could be performed safely. Nanoworms are composed of iron oxide cores that are approved by the US Food and Drug Administration for use in humans (e.g., Feridex); we further examined whether fibrotic livers could be sensitized to nanomaterial toxicity. To investigate nanomaterial safety, we administered peptide-nanoworms (1 mg per kg body weight) or PBS weekly (day 0, 7 and 14) to mice fed DDC or control chow for 3 consecutive weeks (Supplementary Fig. 7a) and found that nanoworms did not exacerbate fibrosis, decrease body weight or induce hepatotoxicity compared to PBS (Supplementary Fig. 7b–e). Serial nanoworm infusions could also introduce experimental artifacts if residual urinary reporters from prior administrations are insufficiently cleared. Analysis of urine samples after the last nanoworm injection (day 14) revealed that both residual fluorescent and mass reporters were cleared within 5 d (day 19) (Supplementary Fig. 7f,g). Collectively, these experiments showed that nanoworms are well tolerated at the dosage selected and require 5 d for full clearance.

We next investigated whether urinary responses are specifically produced by fibrosis-associated proteases such as MMPs by testing urinary sensitivity to pharmacological inhibition of MMPs (Fig. 4a). Whereas infusion of our 10-plex iCORE-encoded nanoworm cocktail (G1–G10) in mice given DDC chow for 3 weeks resulted in a strong increase in ensemble urinary fluorescence over mice given control chow ($P < 0.001$ by analysis of variance (ANOVA); Fig. 4b), urinary responses were significantly attenuated in mice additionally treated with the broad-spectrum MMP inhibitor Marimastat by oral gavage for 2 d before nanoworm administration, resulting in over 70% inhibition of urinary fluorescence ($P < 0.01$; Fig. 4b).

To determine the accessibility of nanoworms to sites of fibrosis, we performed immunofluorescence analysis of liver sections, which revealed that most nanoworms infiltrated freely into the parenchyma and further penetrated periportal zones of active fibrosis, escaping sequestration by resident macrophages (Supplementary Fig. 8a). Compared to control sections, these regions showed substantial upregulation of MMP9, a representative fibrosis-associated MMP (Supplementary Fig. 8b,c). Fibrotic sections treated with DQ gelatin substrates had similar punctate patterns (Supplementary Fig. 8d), confirming the enzymatic activity of collagen-degrading proteases (e.g., MMP2 and MMP9). These results showed that MMPs upregulated during fibrosis are proteolytically active and largely responsible for urinary responses.

We next monitored the processes of fibrosis and resolution by iCORE mass analysis to determine the response of individual biomarkers apart from their collective fluorescence in urine. Mice treated transiently with DDC for 3 weeks and then given DDC-free chow developed distinct fibrosis and resolution windows (0–3 and 7–11 weeks, respectively; Fig. 4c), as verified macroscopically by Sirius red collagen staining of liver sections and hydroxyproline quantification (Fig. 4d,e). With this treatment regime, the amount of liver collagen increased approximately threefold compared to pretreatment amounts after 3 weeks on DDC ($P < 0.005$), persisted from week 3–7 after initial removal of DDC and significantly decreased from week 7–11 ($P < 0.05$ by ANOVA) after sustained DDC withdrawal ($n = 3$). Thus, to monitor the transitions between fibrosing and resolving disease, we administered nanoworms at 0, 3, 7 and 11 weeks into DDC-treated and age-matched control mice and then performed iCORE MS/MS analysis.

The resulting activities of the ten synthetic biomarkers showed markedly divergent kinetics (Fig. 4f). The activities of biomarkers G3 and G4 both strongly increased relative to pretreatment baselines, reaching a plateau by week 11 despite staggered onset at weeks 7 and 3, respectively ($P < 0.01$). G5 and G6 showed opposing kinetics, significantly decreasing ($P < 0.01$) at week 3 before either gradually returning to pretreatment intensities (G5) or persisting to week 11 (G6). G7 tracked with the kinetics of DDC treatment, elevating sharply at week 3 and then rapidly reversing at week 7 ($P < 0.01$ by repeated
measures ANOVA and Tukey’s post test). None of the remaining biomarkers (G1, G2, G8, G9 and G10) deviated from their initial pretreatment activities (n = 10 DDC-treated mice), which was also true for all biomarkers in control mice (Supplementary Fig. 9).

Having identified a set of putative biomarkers for liver fibrosis in the context of DDC, we next sought to crossvalidate promising biomarkers in mice with deletion of Mdr2 (also known as Abcb4), a mechanistically distinct model of liver fibrosis. Crossvalidating biomarker responses in an independent cohort not involved in hypothesis generation is crucial for eliminating potential model-specific artifacts from the use of inbred mice as well as from data overfitting37. Mdr2−/− mice lack a crucial phospholipid transporter that is required for bile stabilization and develop chronic liver injury from birth as a result of bile leakage to the portal tract38.

At 8 weeks of age, Mdr2−/− mice showed evidence of perportal fibrosis as well as significant upregulation of MMP9 compared to age-matched wild-type mice (Fig. 4g and Supplementary Fig. 10a,b). From our library of ten probes, we selected G7 for crossvalidation because it was highly specific for liver fibrosis in our DDC study, tracking with fibrogenesis and declining after fibrotic resolution (Fig. 4f). Similar to our initial observations, we detected significant elevations in urinary fluorescence in 8-week-old Mdr2−/− mice over wild-type mice (Fig. 4h,i). By contrast, G8, a biomarker revealed to be unresponsive in our DDC studies, did not show potential for monitoring fibrosis in Mdr2−/− mice (Fig. 4f). Collectively, these results further corroborated the ability of G7 to track liver fibrosis and underscored the potential for monitoring fibrogenesis with distinct molecular etiologies.

To explore potential improvements in disease classification that could be gained by using more than one biomarker, we further analyzed biomarker responses in our DDC study using receiver operating characteristic (ROC) curves. ROC curves characterize the predictive power of a biomarker by returning the area under the curve (AUC) as a metric, with a baseline AUC of 0.5 representing a random biomarker classifier (Supplementary Figs. 11–13). Within the 0–3 week fibrogenesis window, biomarkers G4, G5, G6 and G7 each discriminated disease with high sensitivity and specificity with associated AUCs ranging from 0.83 to 0.96 (Supplementary Fig. 11), and combinatorial panels, such as the best dual (G5 and G7) and triple (G5, G6 and G7) biomarker combinations, led to improvements in predictivity (0.98 and 1.0, respectively; Supplementary Fig. 13). Conversely, during fibrotic resolution, the ability of candidate biomarkers, such as G1 (AUC = 0.73), to track disease was improved in the dual (G1 and G7) and triple (G1, G7 and G9, AUC = 0.91) biomarker combinations (Supplementary Fig. 13).

Collectively, these experiments demonstrated that liver fibrosis and resolution are revealed by distinct collections of synthetic biomarkers and multiplexed combinations allowed the highest diagnostic performance, illustrating the ability of this platform to noninvasively illuminate otherwise inaccessible aspects of liver disease evolution.

**Early detection of colorectal cancer**

When diagnosed before systemic dissemination, many primary tumors can be treated effectively with conventional clinical interventions39. However, the rates at which most biomarkers are shed from tumors are prohibitively low and cannot be readily augmented40, precluding
Thus there remains a stark mismatch between small tumors detectable by blood biomarkers (>2–5 cm) and the size of tumors that would best respond to treatment (<1–5 mm), resulting in delayed detection, low drug-response rates and reduced overall patient survival. Here we hypothesized that because nanoparticles can passively target tumors to sample proteases through fenestrated angiogenic tumor vessels, cancer-specific proteases could be co-opted to amplify tumor detection through sustained enzymatic release of synthetic urinary biomarkers.

To explore this hypothesis, we compared carcinoembryonic antigen (CEA), a clinically used blood biomarker for colorectal cancer (CRC), with our nanoworms. Because plasma CEA concentrations in patients with CRC are highly variable, we first compiled CEA production rates documented by the American Type Culture Collection from 24 established human CRC lines (14 additional lines were uncharacterized) and found the rates to have a range of well over 4 log units and a median value of 1.65 ng per 10⁶ cells per 10 d (Supplementary Fig. 14). We selected the cell line LS174T to represent colorectal tumors capable of producing CEA near the maximum observed rates (~100× above the median) and validated its ability to secrete CEA in vitro by enzyme-linked immunosorbent assay (ELISA) that had a detection limit of ~0.1 ng ml⁻¹ (Supplementary Fig. 15a,b).

To fully capture the broad spectrum of activities from matrix remodelling proteases shared by most invasive tumors, we infused our biomarker ensemble (G1–G10) into mice bearing LS174T flank tumors (Fig. 5a,b) and detected a significant rise in urinary fluorescence. We verified nanoworm extravasation into the tumor parenchyma by fluorescence imaging of excised tumors and analysis of tissue sections (Supplementary Fig. 16a,b). To test whether urinary responses were specifically produced by proteolysis from MMPs, we treated a separate cohort of tumor-bearing mice with Marimastat for 2 d before G1–G10 administration. Pharmacological inhibition of MMPs resulted in near abrogation of urinary signals, reducing the signal intensities by ~73% (Fig. 5b).

We next directly compared synthetic urinary biomarkers to CEA for early cancer detection. After implantation of LS174T cells, we monitored tumor growth noninvasively by quantifying serum CEA concentrations every 3 d by ELISA, which revealed disease by day 13 when the average tumor volume reached ~330 mm³ (~P < 0.01 by two-way ANOVA, n = 5 mice; Fig. 5c,d). In parallel, we monitored tumor growth by infusing G1–G10 at days 0 and 10 (~130 mm³). Whereas CEA was unable to detect tumor burdens <330 mm³, ensemble urinary responses at day 10 were significantly elevated relative to those in samples before tumor growth, allowing the detection of tumors ~60 % smaller than those detectable with CEA (130 mm³ compared to 330 mm³, respectively; Fig. 5e). To further characterize the discriminatory sensitivity and specificity of these two approaches, we subjected serum CEA and urinary biomarker concentrations to ROC analysis. In contrast to the limited predictive power of CEA for early detection (AUC = 0.61), ensemble urinary fluorescence was highly discriminatory, producing a collective AUC of 0.94 (n = 10 mice; Fig. 5f).

To determine the underlying biomarkers driving the predictivity of the ensemble, the responses of the individual probes were quantified and plotted as ROC curves (Supplementary Figs. 17 and 18). Disease classification by the best-performing individual probes (G1, G2 and G3, AUCs = 0.78–0.81) did not fully recapitulate the multiplexed set but was improved in the dual (G1 and G2, AUC = 0.88) and triple (G1, G2 and G3, AUC = 0.89) biomarker panels (Supplementary Fig. 19). This latter observation underscored the value of using a diverse family of probes for the most sensitive detection.

Having established the potential of biomarker amplification for early cancer detection, we sought crossvalidation in an independent cohort of CRC-bearing mice. In light of the high variability in CEA secretion rates, we hypothesized that our ensemble library of probes could detect tumors that secrete biomarkers at low rates. We selected HCT-15 cells, a genetically distinct CRC line that secretes CEA at significantly reduced rates (Supplementary Fig. 14). As anticipated by the >99% reduction in CEA production relative to LS174T tumors, growth of HCT-15 tumors could not be detected by serum analysis even at up to day 29, when the average tumor burden measured ~1,300 mm³ (Fig. 5c,d), the maximum allowable limit in this mouse model. By contrast, HCT-15 tumors were readily discriminated by nanoworm infusion and urine analysis at day 13 (~150 mm³; Fig. 5c,e), representing, at the minimum, a more than ninefold improvement in detection over CEA (150 mm³ compared to 1,300 mm³, respectively). Collectively, these results showed that synthetic urinary biomarkers have the potential to detect cancer earlier compared to conventional blood biomarkers, with particularly marked enhancements for tumors secreting biomarkers at low rates.

**DISCUSSION**

An ideal biomarker should be secreted at high amounts relative to the native background, remain stable or persistent in circulation until detection, be readily accessible from compositionally simple host...
fluids and discriminate disease with high sensitivity and specificity. In practice, these parameters are often difficult to improve or control for naturally occurring biomarkers and, consequently, many promising biomarkers fail during rigorous evaluation for clinical translation. Here we devised a system of synthetic biomarkers with the capacity to (i) amplify biomarker concentrations through substrate turnover by targeting aberrant protease activities, (ii) release stable, D isomer–enriched mass reporters designed to be present within a narrow mass window free of host molecules, (iii) trigger reporter clearance from blood into urine to reduce matrix complexity and facilitate facile extraction and (iv) simultaneously monitor libraries of candidate synthetic biomarkers in vivo to identify and validate lead biomarkers.

An enabling feature of our platform is the use of a nanoscale scaffold to direct the traffic of peptides in vivo. Although free peptides are typically cleared rapidly from the circulation through urinary secretion, we showed that nanoworm-conjugated peptides are endowed with long circulation times to allow transport into diseased tissues across porous vasculature and are present in urine only after release from nanoworms by disease-associated proteases. Several reports have highlighted the potential of applying peptide substrates to patient serum samples followed by mass spectrometry profiling to uncover disease-specific activity signatures. However, without a delivery mechanism, in vitro serum analysis cannot sample proteases expressed on the membrane of cells residing in the disease microenvironment (e.g., MMP9 expression by liver-resident macrophages in fibrosis). Similarly to blood biomarkers, secreted proteases are markedly diluted in the circulation and are often challenging to detect above highly abundant plasma proteins, potent proteolytic cascades activated during sample collection (e.g., coagulation) and panprotease inhibitors in plasma (e.g., α2-macroglobulin). Here we chose nanoworms as chaperones because iron oxide nanoparticles are safe for use in humans, but a broad range of nontoxic scaffolds, including proteins and sugars (e.g., albumin and dextran, respectively), would also be amenable for peptide delivery. Given the cumulative wealth of nanomaterials, targeting ligands and enhanced delivery strategies available in nanomedicine, we expect this work to be transferrable to many additional formulations to gain access to different organs, types of vasculature and tissue depths.

Our library of isobaric mass tags to track the response of ten peptides in vivo provides a level of multiplexing that is currently challenging to attain with molecular and activity-based imaging probes. The vast majority of these approaches make use of modified protease substrates that emit fluorescent signals after proteolytic cleavage. Consequently, substrate multiplexing is limited by emission overlap as well as the need to emit in the near-infrared window (600–900 nm) to minimize signal attenuation from tissue absorption, constraining most of these studies to single probes. Conversely, our work demonstrates the generation of a synthetic biomarker library that is five to ten times more densely multiplexed than existing state-of-the-art activity-based probes, compares favorably with commercial isobaric tags (e.g., 8-plex iTRAQ) and, with additional parent peptides, is extensible to hundreds of orthogonal mass codes. In addition to its invasiveness, a major limitation of the core biopsy for liver fibrosis is that tissue specimens are only ~1/50,000th the size of an adult liver, leading to sampling variation that can result in inaccurate diagnosis or staging and repeat biopsies. Here we show how nanoparticles accumulate uniformly in the liver, penetrating without bias into regions of active fibrosis to release urinary biomarkers as integrated measures of disease burden. Our work in two models of fibrosis with different mechanisms of induction (i.e., xenobiotic compared to genetic) indicates the value of biomarker G7 for monitoring fibrosis. These results are reflective of fibrosis as a conserved tissue response to diverse chronic liver diseases (e.g., viral hepatitis, alcohol abuse or fatty liver disease) and suggest that biomarker G7 could be useful for monitoring fibrosis stemming from distinct underlying pathologies. Moving forward, an important area for future study will be elucidating the biological mechanisms that are ultimately responsible for the release of individual reporters. This could be accomplished, for example, by comparing urinary signatures from mice lacking specific proteases (e.g., Mmp9−/− mice) to those of their wild-type counterparts or the use of clonodrine liposomes to deplete liver macrophages to identify reporters that track with cellular inflammation.

A major factor preventing early detection of cancer is the tremendous dilution biomarkers experience on release from tumor cells into systemic circulation. Recent computational estimates revealed that solid tumors could potentially remain undetectable for 10–12 years and reach spherical diameters >2.5 cm before biomarker concentrations become sufficiently elevated to indicate disease. The advantage of our system is the ability to amplify tumor responses by leveraging enzymatic turnover (i.e., a single copy of a protease can cleave hundreds of peptide substrates per hour) and the renal system’s natural capacity to remove and concentrate plasma peptides into urine (i.e., from ~5 l of blood to 300 ml void volume). Our study shows that the combined effects of protease amplification and renal concentration can lead to promising results, such as the detection of small tumors that CEA could not discriminate even at the highest tumor burdens allowable in our mouse models. As many tumors do not secrete biomarkers at rates sufficient for detection (or at all), targeting tumor proteases should allow a broader range of cancers to be discovered at an early stage because proteases are uniformly implicated during tumor invasion and metastasis. Extension of this platform to multiple types of cancers would benefit from the development of cancer type–specific tests that could be accomplished by identifying unique biomarker panels for each cancer. Conversely, a highly sensitive, pan-cancer test comprised of a single set of diverse probes would be useful in clinical settings when the primary tumor is already known, such as monitoring for recurrence or metastases after surgical resection of primary tumors.

The successful translation of this platform to humans would require further confirmation of our lead biomarkers in patients as well as quantification of the potential benefits of monitoring biomarker panels compared to single markers. A crucial limitation of existing single-biomarker assays is their relatively poor disease specificity (e.g., CEA is elevated in smokers). These assays can be improved by multiplexing (e.g., prenatal triple screening) or specifying their use in well-defined clinical contexts (e.g., prostate-specific antigen is now recommended for recurrence monitoring but not screening). Similarly, the ability of this platform to differentiate protease-driven diseases (e.g., inflammation compared to cancer) would benefit from multiplexing and serial measurements in high-risk populations.

A general concern with rodent studies is the relatively small number of animals used for hypotheses testing and validation. In this study, the number of mice selected per experimental condition (n = 5–10) allowed reasonable estimation of the mean and variance on the basis of a normal distribution. The statistical power of our study was further bolstered by the prominent effect sizes (e.g., large AUCs) in both diseases studied, as well as the concordant biomarker responses across distinct models. Nonetheless, looking forward and in light of recent expert recommendations regarding biomarker qualifications, the results of this study will require further confirmation and rigorous evaluation in humans.
In summary, this study provides a framework for engineering diagnostic agents that can exploit fundamental features of human disease and physiology for noninvasive urinary monitoring. Future expansion and inclusion of additional enzymatic families (e.g., lipases, nucleases or glycosidases), organ-specific delivery strategies and broader multiplexing capabilities will provide opportunities for systems-level monitoring of disease and elucidating multi-enzymatic networks in health and disease.

METHODS

Methods and any associated references are available in the online version of the paper.

Note: Supplementary information is available in the online version of the paper.

AUTHOR CONTRIBUTIONS

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COMPETING FINANCIAL INTERESTS

The authors declare no competing financial interests.

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1. Sawyers, C.L. The cancer biomarker problem. Nature 452, 548–552 (2008).
2. Hanash, S.M., Pitteri, S.J. & Faca, V.M. Mining the plasma proteome for cancer biomarkers. Nature 452, 571–579 (2008).
3. Sreekumar, A. et al. Metabolomic profiles delineate potential role for sarcosine in prostate cancer progression. Nature 457, 910–914 (2009).
4. Findeisen, P. & Neumaier, M. Functional protease profiling for diagnosis of malignant disease. Proteomics Clin. Appl. 6, 60–78 (2012).
5. Surivova, S. et al. On the development of plasma protein biomarkers. J. Proteome Res. 10, 5–16 (2011).
6. Schwarzenbach, H., Hoon, D.S.B. & Pantel, K. Cell-free nucleic acids as biomarkers in cancer patients. Nat. Rev. Cancer 11, 426–437 (2011).
7. Moon, P.-G., You, S., Lee, J.-E., Hwang, D. & Baek, M.-C. Urinary exosomes and proteomics. Mass Spectrom. Rev. 30, 1185–1202 (2011).
8. Nagrath, S. et al. Isolation of rare circulating tumour cells in cancer patients by microchip technology. Nature 450, 1235–1239 (2007).
9. Lutz, A.M., Willmann, J.K., Cochrane, F.X., Ray, P. & Gambhir, S.S. Cancer screening: a mathematical model relating secreted blood biomarker levels to tumor sizes. PLoS Med. 5, e170 (2008).
10. Haun, J.B. et al. Micro-NMR for rapid molecular analysis of human tumor samples. Sci. Transl. Med. 3, 71ra16 (2011).
11. Edgington, L.E., Verdoes, M. & Bogoy, M. Functional imaging of proteases: recent advances in the design and discovery of substrate-based and activity-based probes. Curr. Opin. Chem. Biol. 15, 798–805 (2011).
12. Nomura, D.K., Dix, M.M. & Crawatt, B.F. Activity-based profiling for biochemical pathway discovery in cancer. Nat. Rev. Cancer 10, 630–638 (2010).
13. Hilderbrand, S.A. & Weissleder, R. Near-infrared fluorescence: application to in vivo molecular imaging. Curr. Opin. Chem. Biol. 14, 71–79 (2010).
14. Bietz, F. & Wisse, E. Structural and functional aspects of liver sinusoidal endothelial cell fenestrae: a review. Comp. Hepatol. 1, 1 (2002).
15. Jain, R.K. & Stylianopoulos, T. Delivering nanomedicine to solid tumors. Nat. Rev. Clin. Oncol. 7, 653–664 (2010).
16. López-Otín, C. & Bond, J.S. Proteases: multifunctional enzymes in life and disease. J. Biol. Chem. 283, 30433–30437 (2008).
17. Schuppan, D. & Afdhal, N.H. Liver cirrhosis. Lancet 371, 838–851 (2008).
18. Horii, S.S. & Gambhir, S.S. Mathematical model identifies blood biomarker-based early cancer detection strategies and limitations. Sci. Transl. Med. 3, 109ra116 (2011).
19. Brenner, C., Tung, C.H. & Weissleder, R. In vivo molecular target assessment of matrix metalloproteinase inhibition. Nat. Med. 7, 743–748 (2001).
20. Kriel, S.J. et al. A unique substrate binding mode discriminates membrane type-1 matrix metalloproteinase from other matrix metalloproteinases. J. Biol. Chem. 277, 23788–23793 (2002).
21. Lutolf, M.P. et al. Repair of bone defects using synthetic micromotors of collagenous extracellular matrices. Nat. Biotechnol. 21, 513–518 (2003).
22. Mahmood, U. & Weissleder, R. Near-infrared optical imaging of proteases in cancer. Mol. Cancer Ther. 2, 489–496 (2003).
23. Turk, B.E., Huang, L.L., Piri, E.T. & Cantley, L.C. Determination of protease cleavage site motifs using mixture-based oriented peptide libraries. Nat. Biotechnol. 19, 661–667 (2001).
24. Park, J.-H. et al. Systematic surface engineering of magnetic nanoworms for in vivo tumor targeting. Small 5, 694–700 (2009).
25. Fickert, P. et al. A new xenobiotic-induced mouse model of sclerosing cholangitis and biliary fibrosis. Am. J. Pathol. 171, 525–536 (2007).
26. Morris, T.A. et al. Urine and plasma levels of fibronectinpeptide b in patients with deep vein thrombosis and pulmonary embolism. Thromb. Res. 110, 159–165 (2003).
27. Choi, H.S. et al. Renal clearance of quantum dots. Nat. Biotechnol. 25, 1165–1170 (2007).
28. Park, J.-H. et al. Magnetic iron oxide nanoworms for tumor targeting and imaging. Adv. Mater. 20, 1630–1635 (2008).
29. Vilanueva, J. et al. Differential extraprotease activities confer tumor-specific peptide signatures. J. Clin. Invest. 116, 271–284 (2006).
30. Vilanueva, J. et al. A sequence-specific exopeptidase activity test (sset) for “functional” biomarker discovery. Mol. Cell. Proteomics 7, S09–S18 (2008).
31. Brown, B.B., Wagner, D.S. & Gysen, H.M. A single-bead decode strategy using electrospray ionization mass spectrometry and a new photoaffinity linker: 3-amino-3-(2-nitrophenyl)propionic acid. Mol. Divers. 1, 1–12 (1995).
32. Ross, P.L. et al. Multiplexed protein quantitation in Saccharomyces cerevisiae using amine-reactive isobaric tagging reagents.ralph.
33. Thompson, A. et al. Tandem mass tags: a novel quantification strategy for comparative analysis of complex protein mixtures by MS/MS. Anal. Chem. 75, 1895–1904 (2003); erratum 75, 4942 (2003); erratum 78, 4235 (2006).
34. Rockey, D.C. et al. Liver biopsy. Hepatology 49, 1017–1044 (2009).
35. Popov, Y. & Schuppan, D. Targeting liver fibrosis: strategies for development and validation of antifibrotic therapies. Hepatology 50, 1294–1306 (2009).
36. Bissing, P., Dargère, D. & Pessayre, D. Assessing variability of liver fibrosis in chronic hepatitis C. Hepatology 38, 1449–1457 (2003).
37. Mishak, H. et al. Recommendations for biomarker identification and qualification in clinical proteomics. Sci. Transl. Med. 2, 46ps42 (2010).
38. Popov, Y., Patsenker, E., Fickert, P., Trauner, M. & Schuppan, D. Mdr2 knockout mice spontaneously develop severe biliary fibrosis via massive dysregulation of pro- and antifibrogenic genes. J. Hepatol. 43, 1045–1054 (2005).
39. Elzioni, R. et al. The case for early detection. Nat. Rev. Cancer 3, 243–252 (2003).
40. D’Souza, A.L. et al. A strategy for blood biomarker amplification and localization using ultrasound. Proc. Natl. Acad. Sci. USA 106, 17152–17157 (2009).
41. Dekker, L.M. et al. Differential expression of protease activity in serum samples of prostate carcinoma patients with metastases. Proteomics 10, 2348–2358 (2010).
42. Russo-Lahit, E., Bhatia, S.N. & Sailer, M.J. Targeting of drugs and nanoparticles to tumors. J. Cell. Biol. 188, 759–768 (2010).
43. Sugahara, K.N. et al. Coadministration of a tumor-penetrating peptide enhances the efficacy of cancer drugs. Science 328, 1031–1035 (2010).
44. Kulasingam, V., Pavlou, M.P. & Diamandis, E.P. Integrating high-throughput technologies in the quest for effective biomarkers for ovarian cancer. Nat. Rev. Cancer 10, 371–378 (2010).
ONLINE METHODS

Nanomaterial synthesis. Nanworms were synthesized according to previously published protocols. Peptides were synthesized at MIT (Swanson Biotechnology Center); isotopically labeled Fmoc amino acids were purchased from Cambridge Isotopes, and 3-Nε-Fmoc-amino-3-(2-nitrophenyl)propionic acid was purchased from Advanced Chemtech. Amine-terminated nanoworms were first reacted with VivoTag 680 (PerkinElmer) to enable in vivo imaging and then with succinimidyl iodoacetate (Pierce) to introduce sulfhydryl-reactive handles. Cysteine peptides and polyethylene glycol–SH were then mixed with nanoworms overnight at room temperature (95:20:1 molar ratio), and excess peptides were removed by size filtration. Peptide-nanoworm stock solutions were stored in PBS at 4 °C.

In vitro protease assays. For substrate screening, FL-peptide-nanworms (2.5 μM by peptide) were mixed with recombinant MMP2, MMP8 and MMP9 (R&D Systems), MMP7 and MMP14 (AnaSpec), thrombin, tissue factor, FXα or cathepsin B (Haematologic Technologies) in a 96-well plate at 37 °C in activity buffers according to the manufacturer’s instructions and monitored with a microplate reader (SpectroMax Gemini EM). For mass spectrometry analysis, equimolar iCORE-encoded nanoworms (Table 1) were incubated with proteases for 2.5 h at 37 °C. Cleavage fragments were purified from nanoworms by size filtration before UV treatment (365 nm; CL–1000 UV crosslinker, UVP). Reporters were then dried by speed vacuum centrifuge and stored at 4 °C.

In vivo imaging. All animal work was approved by the committee on animal care (MIT, protocol 0408–038–11). FVB/Nj mice (Jackson Labs) were fed with 0.1% (w/v) DDC (Sigma) rodent chow for 3 weeks (Research Diets). Fibrotic and age-matched control female mice were i.v. infused with VivoTag 680 (PerkinElmer) to enable in vivo imaging and then with succinimidyl iodoacetate (Pierce) to introduce sulfhydryl-reactive handles. Cysteine peptides and polyethylene glycol–SH were then mixed with nanoworms overnight at room temperature (95:20:1 molar ratio), and excess peptides were removed by size filtration. Peptide-nanoworm stock solutions were stored in PBS at 4 °C.

Characterization of models. For in situ zymography, fibrotic sections were covered with 90 μl solution of 0.5% (wt/vol) low-melt agarose (Sigma) in MMP activation buffer (50 mM Tris, 150 mM NaCl, 5 mM CaCl2 and 0.025% Brij 35, pH 7.5) with 10 μl of DQ gelatin (1 mg ml−1; Invitrogen) and Hoechst dye at 37 °C. Slides were solidified at 4 °C and then incubated at room temperature overnight to promote gelatin proteolysis by tissue proteases. To quantify hepatic collagen, tissue from the right and left lobes (250–300 mg) was hydrolyzed in 5 ml of 6 N HCl at 110 °C for 16 h followed by hydroxyproline quantification as previously described. To quantify CEA, blood was collected from tumor-bearing mice into Capiject microtubes (Terumo) to isolate serum before ELISA (Calbiotech). For immunofluorescence analysis, equimolar nanoworm cocktails (5 μM per peptide) were administered in fibrotic FVB/Nj or tumor-bearing nude mice. After perfusion, livers or tumors were fixed in 4% paraformaldehyde, frozen for sectioning and stained for F4/80 (AbD Serotec), MMP9 (R&D Biosystems), CD31 (Santa Cruz Biotechnologies) and/or FITC (Genetex) before being analyzed by fluorescence microscopy (Nikon Eclipse Ti).

Collection of urinary peptides. Mice were i.v. infused with 200 μl PBS containing equimolar nanoworm cocktails (5 μM per peptide) with EDTA-free protease inhibitor tablets (Roche) to isolate MMP activity. Marinostat was dosed at 100 mg per kg body weight in 0.45% methylcellulose twice daily by orogastric gavage for 2 d before nanoworm infusion. Mice were placed over 96-well plates surrounded by cylindrical sleeves for urine collection. To prevent further reporter degradation, voided samples were spiked with EDTA plus complete protease inhibitors (Roche) immediately after collection. To quantify urinary fluorescence, 2 μl of each sample was incubated with magnetic beads (Dynal) coated with FcITC antibodies (Genetex) in 50 μl binding buffer (100 nM NH4OAc and 0.01% CHAPS) for 1 h at 37 °C, washed twice with 100 mM NH4OAc and eluted twice with 15 μl 5% acetic acid. Samples were neutralized with 2 M Tris and quantified by microplate fluorimetry. For iCORE analysis, samples were irradiated with UV light for 30 min before trichloroacetic acid precipitation (20% final volume) to remove proteins. Soluble fractions were applied to C18 reverse-phase columns (Nest Group) and eluted by step gradients of 20% acetonitrile (ACN) increments in 0.1% formic acid. Sixty percent ACN fractions containing Glu-fib peptides were collected and dried by vacuum centrifuge.

LC-MS/MS analyses. Peptide samples were reconstituted in 5% ACN and 0.1% formic acid and analyzed at MIT or the Taplin mass spectrometry facility (Harvard Medical School). At MIT, peptides were captured and eluted from a C18 nanoflow high-performance liquid chromatography (HPLC) column (75 μm internal diameter; Magic C18 AQ, Michrom BioResources) at a flow rate of 300 nl per min using a water-acetonitrile solvent system with 0.1% formic acid. Electrospray ionization mass spectrometry was carried out on a QSTAR Elite Q-TOF mass spectrometer (AB Sciex). At Harvard, samples were reconstituted in 2.5% ACN and 0.1% formic acid. Samples were injected using a Famos autosampler (LC Packings) into an Agilent 1100 HPLC before mass analysis on a LTQ-Orbitrap (Thermo Electron). To account for discrepancies in urine volumes and concentrations, peak intensities of individual reporters were scaled relative to their respective total iCORE ion currents before normalization against control samples to account for technical and age-related variations.

Statistical analyses. Pearson’s correlation coefficients between different protease profiles were calculated with MATLAB. ANOVA analyses were calculated with GraphPad 5.0 (Prism). For ROC analyses, risk score functions were first estimated by logistic regression on individual biomarkers and then ROC curve analyses of single or biomarker combinations were performed (SigmaPlot).