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**Compact, Light-Weight and Filter-Free: An All-Si
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Compact, Light-Weight and Filter-Free: An All-Si Microspectrometer Chip for Visible Light Spectroscopy

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ABSTRACT

Spectroscopic optical measurements of visible light provide important real-time information in many research and industrial settings. Conventional spectrometers are well established in this context due to their accuracy and reliability, however they have not found widespread use in field-deployed applications due to their bulk (size and weight) and susceptibility to mechanical shock. Here we demonstrate a compact, lightweight microspectrometer chip-scale solution based upon doped silicon nanophotonic elements, known as waveguide array (WGA) pixels. Each pixel is an array of 2.7 μm tall vertically oriented slab waveguides with a subwavelength array period, etched into a P-I-N doped silicon substrate. By adjusting the nanometric dimensions of each element we imbue our pixels with a structural color and each pixel's responsivity spectrum with distinct features. In doing this we forgo the need for the optical elements, such as gratings or interferometers, that one would normally need for a spectrometer. This opens the possibility for dramatic miniaturization and for improvements in mechanical robustness. We fabricate

1 and characterize 144 WGA pixels and use supervised machine learning techniques to perform
2 spectroscopic measurements, including on biological samples, with our 6 mm x 6.5 mm chip. We find
3 that our structurally-colored microspectrometer chip offers performance comparable to that of commercial
4 spectrometer, making it well suited to field deployment in hand-held devices.
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11 **KEYWORDS:** microspectrometers, waveguides, reconstructive spectroscopy, photodetectors,
12 nanofabrication
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18 INTRODUCTION

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21 Medical diagnostics, chemical classification, industrial and environmental monitoring, and scientific
22 research all use spectral information. Typically this is measured by instruments such as grating
23 spectrographs and Fourier transform infrared (FTIR) spectrometers,¹ which combine precisely aligned
24 conventional optical elements such as diffraction gratings or Michelson interferometers with a
25 photodetector or with photodetector arrays. While this work well for many applications, there is currently
26 a trend toward miniaturized optical systems. Indeed, research has recently focused on methods to reduce
27 the weight and size of spectrometers, to better suit the emerging needs of hand-held consumer electronics,
28 sensor networks that are part of the Internet-of-Things, and remote sensing platforms such as un-crewed
29 aerial vehicles (UAVs). Developments in both machine learning and nanophotonics have driven these
30 recent microspectrometer designs, with solutions falling into two camps: the planar filter-array detector-
31 array (FADA) system or the filter-free chip-scale system.^{2,3} Both types rely upon spectral reconstruction
32 via an algorithm as opposed to direct intensity read-out or fast Fourier transform. Prior art in FADA
33 microspectrometer designs include thin film interference and plasmonic filters^{4,5}, colloidal quantum dot
34 absorptive filters incorporated onto charge coupled devices,⁶ photonic crystals filters coupled to CMOS
35 sensors,⁷ and mid-infrared plasmonic filters designed to operate with microbolometer arrays.^{8, 9, 10, 11}
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37 Filter-free microspectrometers on the other hand combine the spectral filtering component and the photo-

1 detecting component that all spectrometers require into a single element. This has been achieved by either
2 spatially varying the chemical composition of semiconductor nanowires to alter the energy bandgap along
3 the nanowire¹², or by creating a set of photodetectors with structural color, which arises due to optical
4 resonances within the nanostructured element.^{13, 14, 15, 16} Our microspectrometer presented here is of the
5 latter type. Previous examples of structurally colored photodetectors include silicon or germanium
6 nanowire arrays with transparent conducting top contacts made of indium tin oxide (ITO)¹⁷ or graphene,¹⁸
7 including a recent silicon nanowire microspectrometer chip with 24 vertical nanowire pixels.¹³ An
8 alternate structurally colored silicon-based example is a microspectrometer using a set of 20 fishnet
9 photodetector pixels which do not require an additional transparent conducting top contact.¹⁴

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21 Our chip, which we denote FIN144, consists 144 waveguide array (WGA) pixels. Each WGA is an array
22 of vertically oriented slab waveguides, each of width s and period Λ . The geometric parameters (s, Λ) take
23 sub- or near-wavelength values. Symmetric optical modes, both transverse magnetic (TM) and transverse
24 electric (TE), are supported by the vertical waveguide arrays and can be excited directly by normally
25 incident plane waves, without the need for additional momentum matching elements such as gratings or
26 prisms.¹⁹ The WGA are formed by inductively coupled reactive ion etching to the $n+$ layer of a P-I-N-I-
27 P doped silicon substrate. Figure 1a shows a schematic of a WGA detector pixel, with the silicon doping
28 indicated by color. Electrical contacts are made to the top $p+$ layer and to the $n+$ layer, forming a vertical
29 P-I-N photodiode, with a responsivity spectrum dictated by the absorption that occurs within the lightly
30 doped ($n-$) intrinsic region for the optical modes supported by the WGA. In each pixel, the spectral
31 filtering and photon detection functions (which are needed in any spectrometer) occur within the same
32 nanophotonic element, allowing for a large reduction in size compared to conventional grating
33 spectrometers. Each WGA is connected to a pad mesa, a $100 \mu\text{m} \times 100 \mu\text{m} \times 2.7 \mu\text{m}$ ($L \times W \times H$) block of
34 unetched silicon, contiguous with each waveguide, that allows us to form the top contact with each pixel
35 and forgoes the need for planarization or a transparent conductive layer. Our chip design consists of 144
36 WGA pixels, organized into a 12×12 array, where each row of 12 pixels share a common electrical back
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1 contact (the $n+$ layer) and each column of 12 pixels share a common top contact (the upper $p+$ layer
2 connected via aluminum wires). Each pixel is 200 μm from its nearest neighbor. We term this
3 configuration as ‘row-and-column addressing’. It allows us to use many more pixels than the number of
4 pins on our chosen chip carrier. By biasing one row and one column at a time, we can measure the
5 photocurrent from an individual WGA pixel. Figure 1b shows a cross-sectional schematic of pixels in
6 rows 1 and 2 and column 1 of FIN144. To electronically isolate the contact rows from one another,
7 trenches are etched through the $n+$ layer and the lower intrinsic layer (thickness: 4 μm) into the lower $p+$
8 layer. We note that this design could be alternatively implemented using a P-I-N doped silicon-on-
9 insulator substrate instead of the P-I-N-I-P substrate.

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21 Following fabrication, which is detailed in the Methods section below, each WGA pixel is characterized
22 by recording its responsivity spectra for TE and TM polarized illumination. Once the pixel responsivity
23 library is recorded, FIN144 is used to measure both narrow and broadband test spectra. This is
24 accomplished by measuring the photocurrents from each pixel created by the unknown spectrum and
25 passing those 144 values to a supervised machine learning algorithm to estimate the unknown spectrum,
26 given the known responsivity of each pixel. We demonstrate that the performance of FIN144 compares
27 favorably with our much larger commercial spectrometer, even for test spectra comprising light
28 transmitted through a bamboo leaf. We demonstrate that the FIN144 device can be used to measure
29 changes in these spectra as the leaf ages over the course of a week. We conclude this paper with a summary
30 of key results and possible applications.

31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 **RESULTS AND DISCUSSION**

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50 The nanofabrication steps used to realize the FIN144 microspectrometer chip are outlined in the Methods
51 section below. The 144 WGA pixels have waveguide widths ranging from 150 nm to 200 nm in steps of
52 10 nm and array periods from 300 nm to 587.5 nm in steps of 12.5 nm. Figure 1c shows a scanning electron
53 micrograph (SEM) of an example WGA pixel with geometric parameters ($s=180$ nm, $\Lambda=425$ nm) and the
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attached pad mesa. In the inset image we see that the waveguide sidewalls are smooth and vertical, which is important for the optical properties of the pixel, reducing scattering losses and ensuring a well-defined set of waveguide modes.²⁰ Figure 1d shows an optical micrograph of one quadrant of FIN144, containing mostly yellow, orange and red pixels. Here the SU-8 bridges connect the WGA pixels of each column, allowing allows each column interconnect wire to span the 6.7 μm deep trenches separating the pixel rows. Figure 1e is an optical micrograph of a single WGA pixel, showing the pad mesa and the column interconnect wire carried atop the SU-8 bridges. The active area of the WGA pixel is the 100 μm x 100 μm red region to the right of the pad mesa. Note that the Al column wire makes direct (ohmic) contact with the top p+ layer of the pad mesa. To allow this, the Al_2O_3 surface passivation layer under the wire's track is removed prior to metallization. A photograph of the entire FIN144 chip prior to wire bonding, including row pads (left) and column pads (right) and their electrical isolation trenches, is provided as Figure 1f. In this image, each bright square comprises diffracted light from a WGA pixel. The chip has an extent of 6 mm x 6.5 mm, which is mostly due to the space required for our row-and-column implementation.

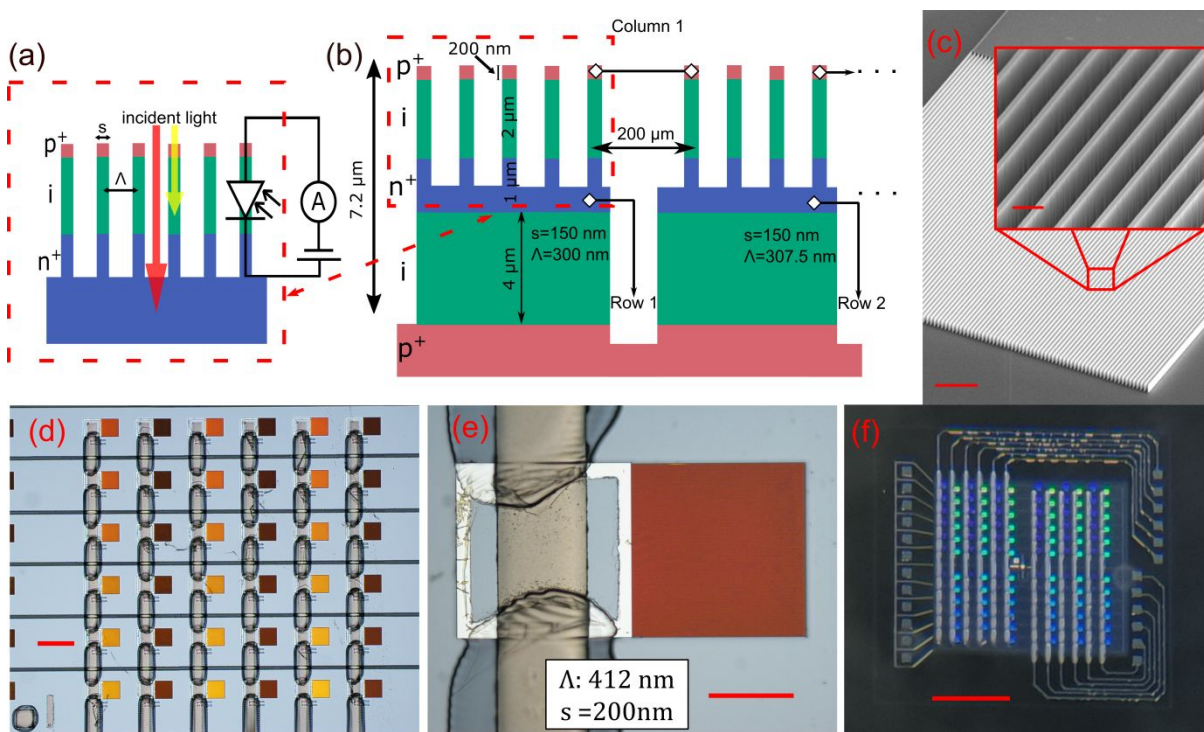


Figure 1. (a) Schematic of WGA pixel, with p+, i(n-) and n+-doped regions shown in red, green and blue, respectively. **(b)** Cross-sectional schematic of two WGA pixels (pixel row 1 and 2) within the same electrically connected column of FIN144 **(c)** SEM of a WGA pixel (scale bar: 17 μm). Inset: zoom-in of WGA (scale bar: 0.5 μm). **(d)** Optical micrograph of upper right quadrant of FIN144, showing WGA pixels, pad mesas, row trenches and SU-8 column bridges (scale bar: 200 μm). **(e)** Optical micrograph of WGA pixel and column wire after metallization (scale bar: 50 μm). **(f)** Photograph of FIN144 microspectrometer chip prior to packaging and wire-bonding (scale bar: 1500 μm).

After packaging FIN144 in a 28-pin dual in-line package and wire bonding it, we perform electrical and photonic characterization. We first measure the current-voltage (I - V) curves for each pixel on the chip. This is done with no illumination (i.e. dark), with TE-polarized illumination (power: 18 μW , $\lambda = 560$ nm) and with TM-polarized illumination (power: 18 μW , $\lambda = 560$ nm). Figure 2a shows the I - V curves for the WGA pixel with geometric parameters ($s=180$ nm, $\Lambda=425$ nm). In the inset, we provide a schematic defining TE and TM polarizations. It can be seen that the I - V curves exhibit the rectifying behavior that one would expect from P-I-N photodiodes, with a reverse saturation dark current of ~ 30 pA. When FIN144 is illuminated (at power: 18 μW , $\lambda = 560$ nm), the current is ~ 1 μA , which represents a $\sim 10^5$ -fold increase over the dark current at -1V bias. At zero bias voltage the light on:off current ratio is around 10^6 . We also note that the open-circuit voltage shifts by +0.7V when the pixel is illuminated. For this pixel, TM polarized light at $\lambda=560$ nm produces 1.7x the photocurrent of the TE polarized case. This difference in general depends on the values of s , Λ and λ . Figure 2b shows the measured external quantum efficiency (EQE) spectrum for this WGA pixel for both TE and TM polarizations measured under -4.8 V reverse bias. It can be seen that the EQE reaches as high as 0.55, which is achieved at $\lambda=515$ nm for TE light. The TM spectrum shows dual-moded behavior with peaks at $\lambda=528$ nm and $\lambda=647$ nm, with the EQE exceeding 0.38 for both. These EQE values surpass our previous reports^{13, 14}. It has been reported that the internal quantum efficiency of nanostructured silicon PIN photodiodes can be improved by the addition of an ALD Al_2O_3 passivation layer.²¹ It might be the case that the improvement in efficiency we obtain in this work might originate from this, although we do not verify this by control experiments. Figure 2c and 2d show the normalized EQE for 10 WGA pixels of fixed width ($s=180$ nm) and varied period, for TM- and TE-polarized illumination, respectively. Here two TM modes (TM₂ and TM₄) and one TE mode

(TE4) are apparent, as is the red-shift in peak EQE wavelength with increasing array period. Importantly, we see that there are spectral peaks across the visible range and into the near-infrared. Previously demonstrated microspectrometers based on structurally-colored photodetectors, where spectral filtering and electrical signal generation occur within the same photonic element, have relied upon a two stacked-detector configuration to expand their operating wavelength range beyond 600 nm into the near-infrared.¹³
¹⁴ The measured normalized EQE spectra for the remaining pixels can be found in the Supporting Information. Figure 2e shows measured responsivity spectra for the same WGA pixels and polarization as Figure 2c, with each successive curve offset by 0.17 A/W. These pixels have peak responsivities up to 0.25 A/W and the gradual change in peak position for both modes (TM2 and TM4) with increasing array period can be seen. To gain physical insight into this, one can analytically model a WGA as a high-contrast grating (HCG). The TE mode cut-off wavelengths, λ_c , can be calculated by solving Equation 1:

$$n_{Si} \tan\left(\frac{\pi n_{Si} s}{\lambda_c}\right) = \tan\left(\frac{\pi(\Lambda - s)}{\lambda_c}\right) \quad (1)$$

where n_{Si} is the refractive index of silicon. For TM modes, we use the same expression, but multiply the right-hand side of Equation 1 by n_{Si}^2 .¹⁹ It has been previously been shown that WGAs have optical absorption spectra with peak wavelengths coinciding with the mode cut-off wavelengths of a HCG, and that the fractional absorption at these peak wavelengths approaches unity.¹⁴ Figure 2f shows that for $s = 180$ nm wide WGA pixels, the analytically calculated TE4 and TM4 cut-off wavelengths and the measured responsivity spectra peak wavelengths (for both TM and TE polarization) are in good agreement. In the following section we demonstrate that although the spectral features of the responsivities are broad in nature, with full-widths-at-half-maxima (FWHM) around 300 nm or more, it is in fact possible to reliably measure narrowband spectra across the visible range with FIN144. This is a distinct feature of the spectral reconstruction approach to microspectrometer design.^{3, 22, 23, 24}

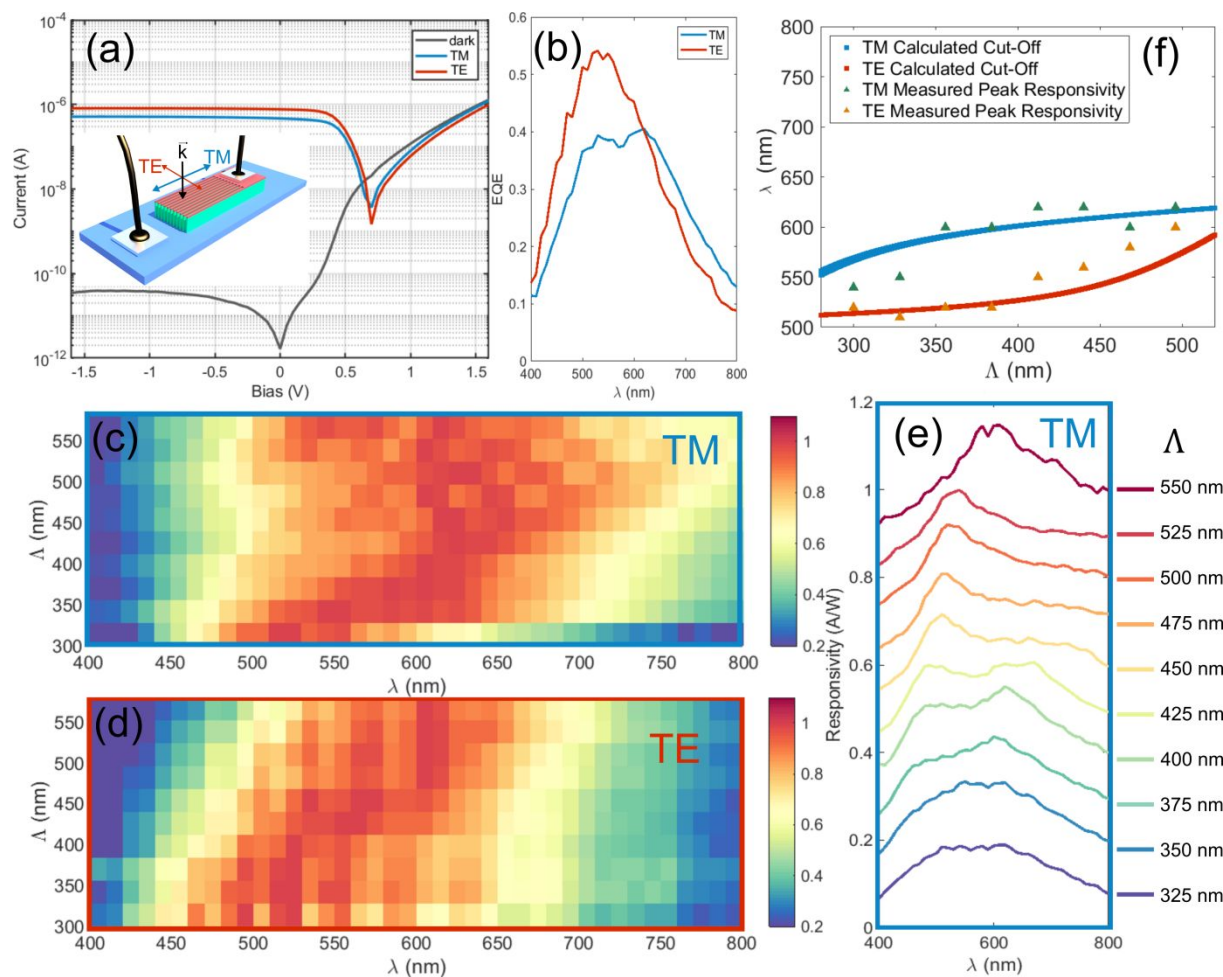


Figure 2. (a) Measured current-voltage curves of WGA pixel ($s=180$ nm, $\Lambda=425$ nm) in the dark (black line) and illuminated with T- (red) and TM (blue) polarized illumination (power: $18 \mu\text{W}$, wavelength = 560 nm). Inset schematically illustrates definitions of TE and TM (b) Measured EQE spectra for TE (red) and TM (blue) illumination. (c)&(d) Measured normalized EQE spectra vs array period for 10 WGA pixels of width 180 nm under TM and TE polarized light respectively. (e) Measured responsivity spectra $s=180$ nm WGA pixels for 10 array periods (successive curves offset by 0.17 A/W for clarity). (f) Analytically calculated mode cut-off wavelengths, with $s=180$ nm, for TE₄ (red) and TM₄ (blue) WGA modes and measured peak responsivity wavelengths for TE (orange) and TM (green) polarizations.

Once the responsivity spectrum of each pixel for both polarizations is carefully measured and stored in a library, FIN144 is ready for use as a microspectrometer chip. To demonstrate the performance of our device, we first illuminate the chip with green light from our monochromator ($\lambda=543$ nm, with a FWHM of 4.8 nm). We choose a wavelength that is not used during collection of the responsivity library so as not to unduly bias the resulting spectral reconstruction. This is known as hold-out validation in machine learning. For all the sample reconstructions henceforth, we only consider narrowband spectra not found

1 in the responsivity library. Although our microspectrometer chip is able to measure unpolarized spectra,
2 here we will only use TE polarized light for the following reconstructions, as we find that including
3 photocurrent data from TM polarized light in reconstructions provides only small improvements in
4 accuracy. Figure 3a shows the measured normalized photocurrents from each WGA pixel in FIN144, as
5 well as the expected photocurrents found by multiplying the previously measured responsivities by the
6 measured reference spectrum from our commercial grating spectrometer. We also plot the expected
7 photocurrents as calculated from the measured responsivity library and the reconstructed spectrum from
8 FIN144. All three sets of photocurrents are in good agreement, with only some small discrepancies
9 between expected and actual photocurrents. Figure 3b shows the reference spectrum in question as well
10 as the LASSO regularization-based estimate from FIN144. We find the agreement between the two is
11 excellent in terms of key features such as FWHM and center wavelength, with a mean square error (MSE)
12 of 1.3×10^{-3} . Since the test spectrum has a FWHM of 4.8 nm, we estimate FIN144 to have a spectral
13 resolution (when measuring narrowband spectra) of 2.5 nm, which compares favorably with our reference
14 spectrometer's stated value of 1 nm. Figure 3c shows a set of nine narrowband spectra as measured by
15 FIN144 showing the chip is suitable for use with spectra spanning the visible range. The reference spectra
16 can be found in the Supporting Information. To the best of our knowledge, this is the largest demonstrated
17 operating range (423 nm to 687 nm) of a visible wavelength microspectrometer based on structurally-
18 colored photodetectors.

19 Next, we evaluate the performance of FIN144 when illuminated with a naturally-occurring test spectrum.
20 Here we pass white light through a freshly harvested bamboo leaf and measure photocurrents at each
21 WGA pixel, as shown in the inset of Figure 3d. LASSO regularization is ill-suited to reconstructing
22 broadband spectra,²⁵ so instead we use Tikhonov regularization to estimate the spectral density that
23 minimizes the L_2 norm. Figure 3d shows the fresh bamboo leaf transmission spectrum as measured by
24 FIN144 and by our reference spectrometer. It can be seen that these are in good agreement, with similar
25 peak transmission wavelengths and FWHMs. The MSE between the FIN144 reconstruction and the
26 reference spectrum is 0.0086, indicating a close match between the two as the maximum possible MSE is
27

1 unity. Finally, we test a potential ‘real-world’ application of our device, including confirming the temporal
2 stability of our microspectrometer chip, calibration process and machine learning algorithm. To do this
3 we perform spectral measurements on a freshly harvested bamboo leaf and on the same leaf after one
4 week, without recalibrating FIN144 (that is, using the original responsivity library). Photographs of the
5 fresh and aged leaf are shown as Figure 3e. Figure 3f and 3g show the two spectral measurements,
6 conducted a week apart, from our reference spectrometer and from FIN144 respectively. The MSE
7 between the FIN 144 reconstruction and reference spectrum is 0.0374. These measurements reveal that as
8 the leaf dehydrates, its transmission reduces. The relative amplitudes of the transmission peaks a week
9 apart are 0.18 from the reference spectra and 0.17 from the FIN144 results. This test is an important step
10 towards a field-deployable microspectrometer, where the time evolution of the relative amplitude of
11 spectral features can be an important indicator. For example, it can indicate water loss in vegetation on
12 sections of farmland or rainforest. Furthermore, by comparing to previously demonstrated structural
13 color-based visible-range microspectrometers we note that using 144 pixels (compared to 20-40^{13, 14})
14 gives an improved match to the broadband reference spectra. The microspectrometer results here now are
15 much closer to those of the commercial (macro) spectrometer than previously achieved, while still being
16 many times smaller (our microspectrometer chip is only 6 mm x 6.5 mm x 0.5 mm).
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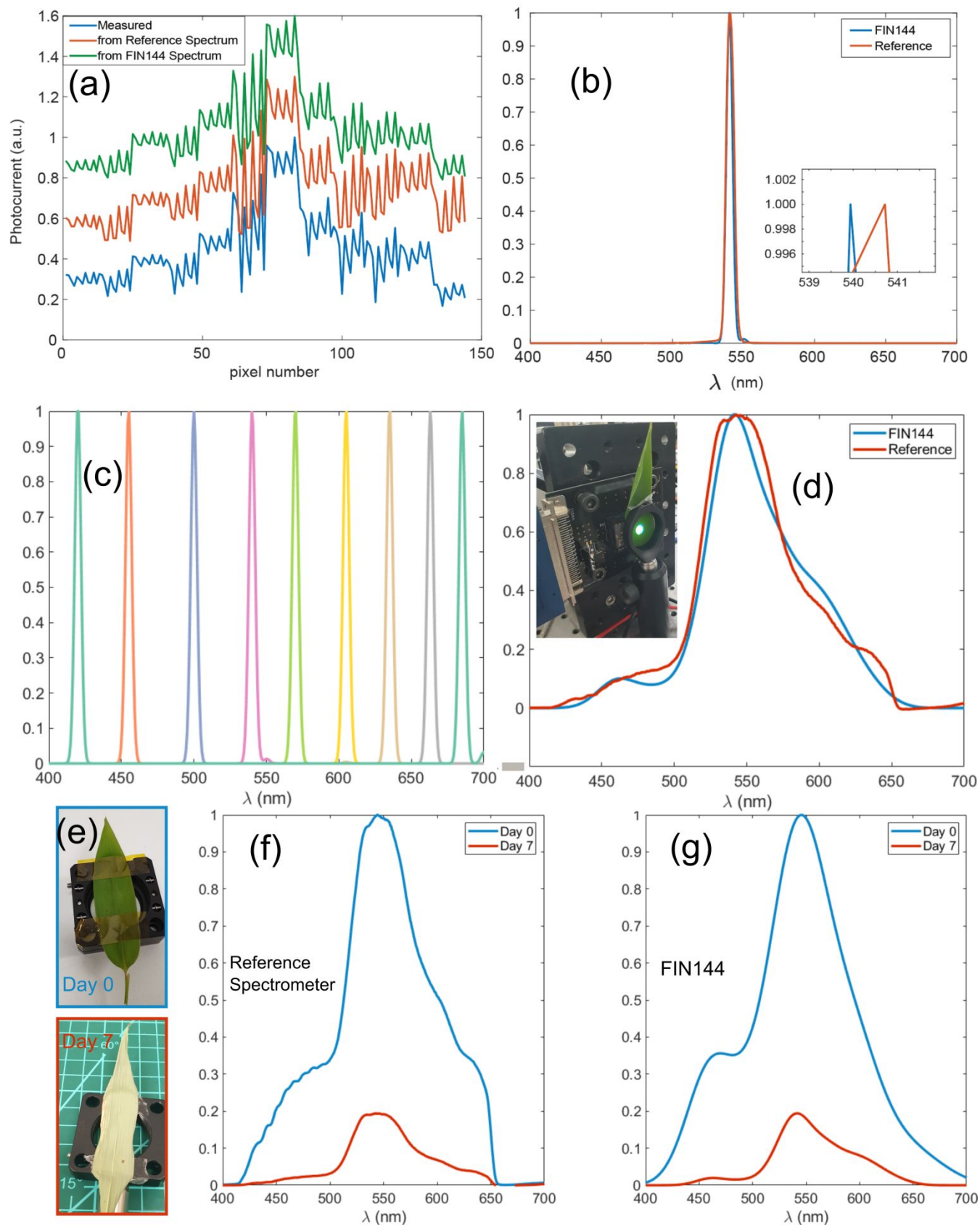


Figure 3. (a) Normalized photocurrents from all 144 WGA pixels as measured (blue line), calculated from reference spectrum (red) and calculated from FIN144 spectrum (green), for $\lambda=543$ nm narrowband test spectrum. Note that for better readability, the photocurrents are offset by 0.3 and 0.6 in the ‘from

1 Reference Spectrum' and 'From FIN144 Spectrum' plots, respectively. **(b)** Reference (red) and
2 reconstructed (blue) narrowband (L1) test spectrum. Inset shows the zoomed view at the peak
3 wavelengths. **(c)** Set of reconstructed narrowband (L1) spectra, spanning the visible range. **(d)**
4 Reconstructed (blue) and reference (red) broadband bamboo leaf transmission spectra. Inset: bamboo leaf
5 placed in front of FIN144 for spectral measurement. **(e)** Bamboo leaves after zero (blue box) and seven
6 (red) days since harvest. **(f)** Reference bamboo leaf transmission spectra after zero (blue) and seven (red)
7 days. **(g)** Bamboo leaf transmission spectra as measured by FIN144 after zero (blue) and seven (red) days.
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11 CONCLUSION

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15 In summary, we demonstrate a visible-range microspectrometer chip comprised of 144 nanophotonic
16 pixels, where each pixel is an array of vertically-oriented, P-I-N-doped silicon waveguides. Each WGA
17 pixel has a unique combination of waveguide period and waveguide widths, meaning that it has a unique
18 responsivity spectrum for both TE and TM polarizations. We show that by characterizing the latter, a
19 responsivity library can be reconstructed that allows our device to function as a microspectrometer. The
20 external quantum efficiencies of these pixels exceed 0.5 and are competitive with commercial silicon P-
21 I-N photodiodes. Once characterized, photocurrent information from each pixel can be used with
22 supervised machine learning to determine the best estimate of the spectrum illuminating the chip. We test
23 the chip as a microspectrometer with both narrowband spectra from a monochromator and broadband
24 spectra from a biological sample (bamboo leaf). We find our microspectrometer performs well in both
25 cases, with an estimated resolution of 2.5 nm and with low MSE values between reference and estimated
26 spectra across the visible range. This is enabled by judicious choice of reconstruction algorithm (i.e. L_1
27 or L_2 regularization). Finally, we demonstrate a potential 'real-world' application of our system by
28 conducting bamboo leaf transmission measurements a week apart without recalibrating the responsivity
29 library or modifying the machine learning algorithm. We find good agreement between our
30 microspectrometer and the reference measurements. Our silicon based structurally-colored
31 microspectrometer will suit spectroscopic applications where size, weight and power consumption are
32 key considerations, for example in hand-held or UAV-mounted devices.
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METHODS

Nanofabrication methods. To fabricate the microspectrometer chip we use the same P-I-N-I-P silicon doping structure used in Ref. ¹⁴ To define each of the 144 waveguide arrays and their pad mesas, electron beam lithography (Vistec EPBG5000+) and electron beam evaporation (IntlVac NanoChrome II) are used to create 25 nm thick Al etch masks, via polymethyl methacrylate (PMMA) lift-off. The masks protect the underlying silicon during the subsequent nano-etching step. To create high-aspect ratio (~120 nm wide, 2.7 μm tall) Si nanofeatures, a pseudo-Bosch inductively coupled plasma-reactive ion etching (ICP-RIE) recipe is used (Oxford PLASMALAB 100), although it may be possible to form these structures using metal-assisted chemical etching as well.²⁶ Here an etch gas chemistry at 10 mTorr of 40 sccm of SF_6 and 90 sccm of C_4F_8 is used with 1200 W ICP power and 38 W forward bias power. This recipe ensures the WGAs have smooth and vertical sidewalls, which is important for achieving the desired electrical and optical properties. We etch the silicon to the embedded $n+$ layer, so that each WGA pixel is a vertically stacked P-I-N diode. Next, direct-write photolithography (Intelligent Micropatterning SF100 XPRESS with AZ-4562 photoresist) is used to define the trenches that separate the WGA pixel rows. The trenches are etched 6.7 μm from the embedded $n+$ layer, all the way through the lower intrinsic region to the $p+$ layer below. This electrically isolates the pixel rows from one another, but pixels within a row now share a common cathode, that is, the $n+$ layer. We also etch trenches to isolate regions of the substrate where the bond pads and leads will be placed. The next step involves the atomic layer deposition (ALD, Fiji F200) of 8 nm Al_2O_3 for surface passivation of the bare silicon, which has been shown to reduce the deleterious effect of surface trap states and improve the external quantum efficiency of nanostructured silicon photodiodes.²¹ Next, photolithography is used to electrically connect the WGA pixels in each column. This is done by forming SU-8 2010 epoxy “bridges” to link the pixels. We under-expose the SU-8 during UV photolithography patterning so that the corners and edges of the SU-8 bridges are rounded out and smoothed. This gives a gradual change in height from the top $p+$ pad of the WGA pixels to the wirebond pads on the $n+$ layer 2.7 μm below, instead of an abrupt step change in height which

1 would be problematic for metallization. The SU-8 bridges permit metallization via DC sputtering (Intlvac
2 NanoChrome) of 250 nm Al followed by lift-off. The metal contacts atop the pad mesas also perform the
3 function of blocking light, thereby preventing the pad mesas from contributing to the measured
4 photocurrent. Finally, the chip is mounted in a DIP-28 chip carrier and each row pad and column pad is
5 wire bonded (F&S Bondtech) to a unique pin in the carrier.
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13 **Characterization and Experimental Methods.** The microspectrometer chip is mounted on a custom
14 printed circuit board (PCB) with each row and column of WGA pixels connected to a single external
15 transimpedance amplifier (SRS SR570) via a 2-channel 16-to-1 analog multiplexer integrated circuit
16 (Analog Devices ADG726). The rows connected to one multiplexing channel, while the columns are
17 connected to the other. The multiplexer is controlled via logic signals from an Arduino Uno connected to
18 a PC running MATLAB. This allows us to rapidly switch the bias voltage to the desired row and column
19 address, enabling us to measure the photocurrent from the desired WGA pixel. We measure the TE- and
20 TM-mode responsivities of each WGA pixel by illuminating it with a known power and wavelength.
21 Illumination is from a laser driven white light source (Energetiq EQ-99X) coupled to a monochromator
22 (Princeton Instruments Acton SP2150). The beam is mechanically chopped (Thorlabs MC2000B) at 186
23 Hz with a 50% duty cycle. The polarization angle is controlled by a linear polarizer and half-waveplate
24 (Thorlabs LPVISE100-A and AHWP05M-580). The photocurrents generated in each pixel at each
25 illumination wavelength under a reverse bias voltage of -4.8 V are measured by a lock-in amplifier (SRS
26 SR830). Photocurrents are collected from all 144 pixels for each wavelength and then stepped to the next
27 wavelength by the monochromator, from 400 nm to 800 nm in steps of 5 nm. The full-width at half-
28 maximum of light from the monochromator is measured to be less than 5 nm. After photocurrent data
29 collection is complete, the microspectrometer chip is replaced with an optical power meter (Thorlabs
30 PM100D, used with a 300 μm pinhole aperture) to determine the intensity spectrum illuminating the chip.
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wavelengths not included in the calibration set (e.g. 573 nm) for narrowband testing. or replaced with a mirror to test broadband transmission spectra, for example, from bamboo leaves. The photocurrents generated at each pixel are measured using the multiplexing PCB, preamplifier and lock-in amplifier. For comparison of results a reference spectrum is measured for each sample using a commercial grating spectrometer (Ocean Optics QEPro).

Supervised Machine Learning Method for Spectral Reconstruction. To reconstruct the spectrum of light incident on the microspectrometer chip, the photocurrents \mathbf{I} measured from the WGA pixels are input into a supervised machine learning algorithm, along with the *a priori* measured responsivity spectra matrix, \mathbf{R} , where \mathbf{R} is a 144 x 81 matrix and \mathbf{I} is a 144 x 1 vector. The reconstructed spectrum, \mathbf{S} , is the solution that minimizes the inverse problem in Equation 2, that is, it is the vector that minimizes the L_γ -norm,

$$\|\mathbf{I} - \mathbf{R}\mathbf{S}\|_2^2 + \epsilon_\gamma \|\mathbf{S}\|_\gamma^\gamma \quad (2) \text{subject to regularizer } \epsilon_\gamma, \text{ where } \gamma = 1 \text{ or } 2. \text{ Clearly, } \mathbf{S} \text{ is a } 81 \times 1 \text{ vector.}$$

We detail how we increase the dimensionality of \mathbf{S} to match that of the reference spectra (517 x 1) in the Supporting Information. For narrowband spectra, such as emission lines, L_1 regularization (which is also known as LASSO regression and is equivalent to a linear support vector machine)²⁷ is used, as it has been previously shown to be well suited to this sparse spectral reconstruction problem.¹³ To find a suitable regularizer, ϵ_1 , we use a geometric sequence of 100 candidate values from $0.01\epsilon_{1max}$ to ϵ_{1max} , where ϵ_{1max} is the largest regularizer that still gives a non-null estimate of \mathbf{S} . 6-fold cross-validation is then used to find an estimate of the spectral reconstruction's mean-square-error (MSE) for each candidate value of ϵ_1 . In this case, each test and training data set corresponds to 24 and 120 WGA pixels, respectively. The value of ϵ_1 with the lowest MSE is then used as the regularizer for the narrowband spectral reconstruction. For broadband (non-sparse) spectra the LASSO method is poorly suited, so instead we use Tikhonov regularization (also known as ridge regression or weight decay) to minimize the L_2 norm. The regularizer for this method, ϵ_2 , is chosen via the L-curve method.²⁸

ASSOCIATED CONTENT

Supporting Information

This material is available free of charge via the internet at <https://pubs.acs.org>

1 Narrowband test spectra; Matching FIN144's dimensionality with the commercial spectrometer;
2 Photodetector dark current measurement; Measured normalized EQE spectra photodetector; Simulated
3 EQE spectra photodetector
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AUTHOR CONTRIBUTIONS

J.J.C. and K.B.C conceived the idea for this work. J.J.C. performed the device design, measurements and the experimental data analysis. J.J.C., J.M., D.W., and V.R.S. performed fabrication. J.J.C. and J.M. designed the experimental set-up. All authors contributed to the writing of the manuscript.

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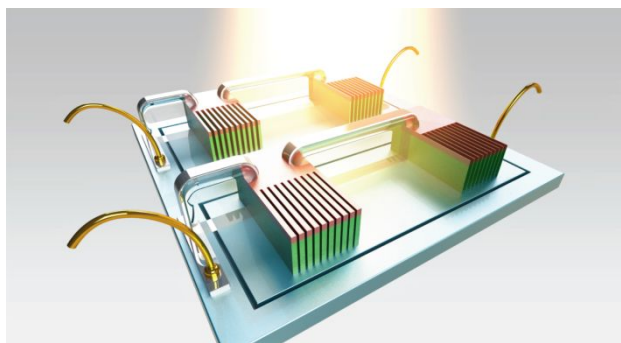
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For Table of Contents Use Only



Title: Compact, Light-Weight and Filter-Free: An All-Si Microspectrometer Chip for Visible Light Spectroscopy

Authors: Jasper J. Cadusch, Jiajun Meng, Dandan Wen, Vivek Raj Shrestha and Kenneth B. Crozier

TOC graphic synopsis:

Left: Schematic illustration of the compact silicon waveguide array microspectrometer chip