## Supplemental Information

# Instrument-free synthesizable fabrication of label-free optical biosensing paper strips for the early detection of infectious keratoconjunctivitides 

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## Raman spectrum of 2-NAT molecule



Figure S1. (A) The molecular structure of 2-naphthalenethiol (2-NAT) and (B) Raman spectra of a $1 \mu \mathrm{M}$ 2-NAT molecule for GNPs deposited on paper with optimal SILAR condition. The prominent Raman peaks of the 2-NAT molecule were obtained at $767 \mathrm{~cm}^{-1}$ ( $\mathrm{C}-\mathrm{H}$ wag), $842 \mathrm{~cm}^{-1}$ ( $\mathrm{C}-\mathrm{H}$ twist), $1064 \mathrm{~cm}^{-1}$ (symmetric C-H bend), $1378 \mathrm{~cm}^{-1}$ (ring stretch), and $1621 \mathrm{~cm}^{-1}$ (ring stretch).

Distribution of GNPs with number of SILAR cycles (10 mM SILAR reagents)


Figure S2. Distributions of SILAR-synthesized GNPs deposited on paper with (A) two, (B) four, (C) six, (D) eight, and (E) 10 SILAR cycles (top: SEM. Scale bar=250 nm; middle: GNP distribution; bottom: schematic distribution). The surface density of GNP was represented by the number of GNP per unit area (no. $/ \mu \mathrm{m}^{-2}$ ). RMSE indicates the root mean square error between the Gaussian-predicted data and experimental data.

## Distribution of SILAR reagent-concentrated GNPs with number of SILAR cycles



Figure S3. Distributions of SILAR-synthesized GNPs deposited on paper with (A) two, (B) four, (C) six, (D) eight, and (E) 10 SILAR using $1 \mathrm{mM} \mathrm{HAuCl}{ }_{4}$ and $\mathrm{NaBH}_{4}$ SILAR reagents. (top: SEM. Scale bar=250 nm ; bottom: GNP distribution). The surface density of GNP was represented by the number of GNP per unit area (no. $/ \mathrm{um}^{-2}$ ). RMSE indicates the root mean square error between the Gaussian-predicted data and experimental data.


Figure S4. Distributions of SILAR-synthesized GNPs deposited on paper with (A) two, (B) four, (C) six, (D) eight, and (E) 10 SILAR using $5 \mathrm{mM} \mathrm{HAuCl} I_{4}$ and $\mathrm{NaBH}_{4}$ SILAR reagents. (top: SEM. Scale bar=250 nm ; bottom: GNP distribution). The surface density of GNP was represented by the number of GNP per unit area ( $\mathrm{no} . / \mathrm{\mu m}^{-2}$ ). RMSE indicates the root mean square error between the Gaussian-predicted data and experimental data.


Figure S5. Distributions of SILAR-synthesized GNPs deposited on paper with (A) two, (B) four, (C) six, (D) eight, and (E) 10 SILAR using $20 \mathrm{mM} \mathrm{SILAR} \mathrm{HAuCl}_{4}$ and $\mathrm{NaBH}_{4}$ reagents. (top: SEM. Scale bar=250 nm; bottom: GNP distribution). The surface density of GNP was represented by the number of GNP per unit area (no. $/ \mathrm{mm}^{-2}$ ). RMSE indicates the root mean square error between the Gaussian-predicted data and experimental data.

## - Computational modeling



Figure S6. (A) Finite element model of GNPs with SILAR cycles and computational results of GNP diameter and interparticle gap distance-dependent LSPR effect with (B) two, (C) four, (D) six, (E) eight, and (F) 10 SILAR cycles on the electromagnetic field (EF).

## Sensitivity



Figure S7. Representative SERS spectra with (A) different 2-NAT concentrations ( $10^{-12} \sim 10^{-6} \mathrm{M}$ ) and (B) low concentrations of 2-NAT on SILAR-synthesized GNPs deposited on SERS paper. Gray indicates a 2-NAT molecule-characterized peak at $1378 \mathrm{~cm}^{-1}$.

## Enhancement factor



Figure S8. Raman spectra of 1 mM 2-NAT on bare paper and 1 pM 2-NAT and no analyte on SILARsynthesized GNP paper strip.

The enhancement factor (EF) was calculated as the difference in Raman intensity between two different substrates as

$$
\begin{equation*}
E F=\left(\frac{I_{\text {SERS }}}{I_{\text {bare }}}\right)\left(\frac{N_{\text {bare }}}{N_{\text {SERS }}}\right) \tag{S1}
\end{equation*}
$$

where $I_{\text {SERS }}$ and $I_{\text {bare }}$ are the Raman intensity of the molecule on the SERS and bare papers, respectively, and $N_{\text {SERS }}$ and $N_{\text {bare }}$ are the average number of adsorbed molecules in the scattering volume for SERS and non-SERS areas, respectively. ${ }^{1}$ Assuming that the probe molecules were uniformly distributed on the substrates, the number of adsorbed molecules can be estimated as

$$
\begin{equation*}
N=\left(N_{A} \cdot c \cdot \frac{V_{\text {droplet }}}{A_{\text {spot }}}\right) A_{\text {laser }} \tag{S2}
\end{equation*}
$$

where $N_{\mathrm{A}}$ is Avogadro's constant, $c$ is the concentration of the probe molecule, $V$ is the volume of the molecule droplet, $A_{\text {spot }}$ is the size of the substrate, and $A_{\text {laser }}$ is the size of the laser spot. ${ }^{2}$ Since the same methods for assessing the Raman measurement were applied to two substrates, the parameters of $N_{\mathrm{A}}$ for $2-\mathrm{NAT}, V, A_{\text {spot }}$, and $A_{\text {laser }}$ were the same.

Therefore, Eq. (S2) can be written as

$$
\begin{equation*}
E F=\left(\frac{I_{\text {SERS }}}{I_{\text {bare }}}\right)\left(\frac{c_{\text {bare }}}{c_{\text {SERS }}}\right) \tag{S3}
\end{equation*}
$$

where $c_{\text {SERS }}$ and $c_{\mathrm{R}}$ are the concentration of 2-NAT molecule on the GNP and bare papers, respectively.

## Bio-applications



Figure S9. PCA-SVM scores ( $n=15$, each): (A) to detect the presence of keratoconjunctivitis using SVM classifier with linear kernel and to classify the normal eye and keratoconjunctivitides using SVM classifier with (B) the second-order polynomial kernel ( $d=2$, Eq. S4), (C) Gaussian kernel ( $\sigma=2$, Eq. S5), and (D) Hilbert transform radial basis function kernel ( $\rho=\mathbf{2}, \boldsymbol{a}=1, \boldsymbol{b}=\mathbf{2}$, Eq. S6). ${ }^{3,4}$

$$
\begin{align*}
& K_{\text {poly }}\left(\mathbf{x}_{1}, \mathbf{x}_{2}\right)=\left(\left\langle\mathbf{x}_{1} \cdot \mathbf{x}_{2}\right\rangle+1\right)^{d}  \tag{S4}\\
& K_{\text {Gaussian }}\left(\mathbf{x}_{1}, \mathbf{x}_{2}\right)=\exp \left(-\left\|\mathbf{x}_{1}-\mathbf{x}_{2}\right\|^{2} / 2 \sigma^{2}\right)  \tag{S5}\\
& K_{\text {HtBRF }}\left(\mathbf{x}_{1}, \mathbf{x}_{2}\right)=\exp \left(-\rho \sum_{i}\left|\mathbf{x}_{1 i}^{a}-\mathbf{x}_{2 i}^{a}\right|^{b}\right) \tag{S6}
\end{align*}
$$

## References

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