



UNIVERSITY OF MINNESOTA

Field-level yield benefits and risk effects of intensive soybean management across the U.S.

IN A BEAN POD...

- ▶ Performance of high- and low-input soybean management systems across the U.S. were evaluated
- ▶ High-input systems increased yield but effect was inconsistent (-4.9 to 12.7% of average yield) among states
- ▶ High-input systems minimally reduced the average cost of yield risk (<3% of average yield)
- ▶ High-input systems do not consistently protect soybean yield from downside yield risk compared to low-input systems

INTRODUCTION

Average crop yields will need to increase during the next three decades to meet expected increases in food demand without a massive cropland area expansion (Tilman et al., 2011; Alexandratos and Bruinsma, 2012; Grassini et al., 2013). This challenge can be met by identifying and adopting best management practices for major production environments. Best management practices are those that, for a given environment, consistently result in high yields with reduced downside yield risk from issues such as adverse weather and pest damage. Farm economic realities also require that these practices be profitable. Hence, farmers regularly explore opportunities to increase yield and profit and to minimize production risks.

At planting time, weather and pest pressure during the growing season are unknown and so uncertainty exists about which specific management practices and inputs will be needed and at what amounts. This uncertainty, coupled with increased soybean prices and commercial marketing, have encouraged soybean farmers to adopt high-input management systems to protect and maximize yield and profitability (Marburger et al., 2016; Orłowski et al., 2016). Such systems involve prophylactic application of multiple inputs such as biological and pesticidal seed treatments, soil and foliar fertilizers, and foliar pesticides regardless of the soil nutrient status or anticipated disease pressure.

Although high crop yields are important, production stability across regions and over time is important. Yield variance is a symmetric measure of variability around the mean, and an important measure of production stability, but unusually low yields are often more problematic than unusually high yields. Hence, a key question is whether the prophylactic application of multiple inputs in high-input management systems better protects from yield loss compared with low-input management systems, i.e., does it lower downside yield risk?

Our objective was to measure the effect of high-input management systems on soybean yield and risk by analyzing yield data from field experiments over the period 2009–2014 across the U.S. (Figure 1). The high-input cropping system included fungicide, insecticide and biological seed treatments, soil and foliar fertilizer and foliar fungicide and insecticide applications. None of these inputs were applied in the low input system. We examined two hypotheses: (i) high-input management systems consistently increase soybean productivity across the U.S. and (ii) high-input management systems are associated with lower yield risk as measured by the variance,



We examined two hypotheses:

(i) high-input management systems consistently increase soybean productivity across the U.S.

and

(ii) high-input management systems are associated with lower yield risk as measured by the variance, skewness, and kurtosis of soybean yield.





Figure 1. Locations and number of site-years within each state between 2009 to 2014.

skewness, and kurtosis of soybean yield. Reduced variance and increased skewness are desirable as they lead to lower risk exposure (from a lower variance) and lower exposure to unfavorable events implied by the lower tail of the yield distribution (higher skewness). Decreased kurtosis means a lower exposure to rare events in the tails of the yield distribution which is desirable. Our analysis documents the extent to which intensive soybean management contributes to high yield by reducing yield loss and risk exposure.

MATERIALS AND METHODS

We used yield data from replicated field experiments over the period 2009–2014 within 10 states (Figure 1). Within each experiment, two cropping systems were applied. The high-input cropping system included fungicide, insecticide and biological seed treatments, soil and foliar fertilizer and foliar fungicide and insecticide applications. None of these inputs were applied in the low input system. All experiments were non-irrigated and the same background management practices (e.g., tillage, previous crop, row spacing, seeding rate) were used for both systems.

Our analysis relies on a moment-based approach by evaluating the mean, variance, skewness, and kurtosis of soybean yield conditional on state (average of all locations in a state) and cropping system (Shi et al., 2013). Mean yield reflects average productivity of each cropping system in each state. The variance captures the variability of soybean yield around its mean. The skewness measures the asymmetry of the yield distribution, with a negative skewness capturing exposure to losses located in the lower tail of the distribution (downside risk). The kurtosis measures the thickness in the tails of the distribution of soybean yield. A large kurtosis is associated with a high risk because it indicates high probabilities of extreme yields (low and high).

RESULTS

High-input management systems increased soybean yield compared to low-input systems within each of the examined states, apart from KS (Figure 2). In these states, the estimated probability of a positive yield difference (high- minus low-input management systems) was $\geq 97\%$. The realized yield increases ranged between 6.3 to 12.7% of average yield, with the largest yield benefits observed in northern states (MN and WI). The only state where high-input management did not increase yield compared to low-input management was KS.

Our analysis shows that high-input management had effects on yield variance that varied by state (Figure 3). In AR, MN, and LA, low-input systems had a greater yield variance than high-input systems, while the opposite was observed in the other

Figure 2. Distribution of the yield difference (bu/ac) between high- and low-input cropping systems in each state and the probability (P) as a percentage that the yield difference > 0 in the posterior sample distribution (n=5,100). Within each state, the red dashed line shows the zero yield difference, and the black line indicates the mean yield difference (Yd).

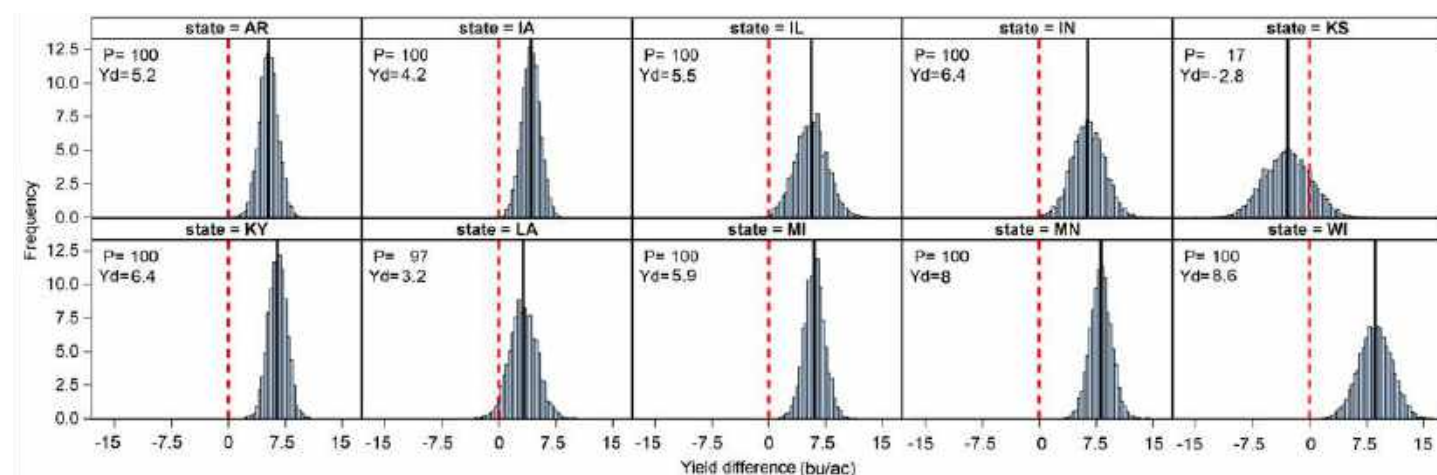


Figure 3. Distribution of yield variance difference between high- and low-input cropping systems in each state and probability (P) as a percentage that the yield variance difference > 0 in the posterior sample distribution (n=5,100). Within each state, the red dashed line shows the zero variance difference, and the black line indicates the mean variance difference (Yd).

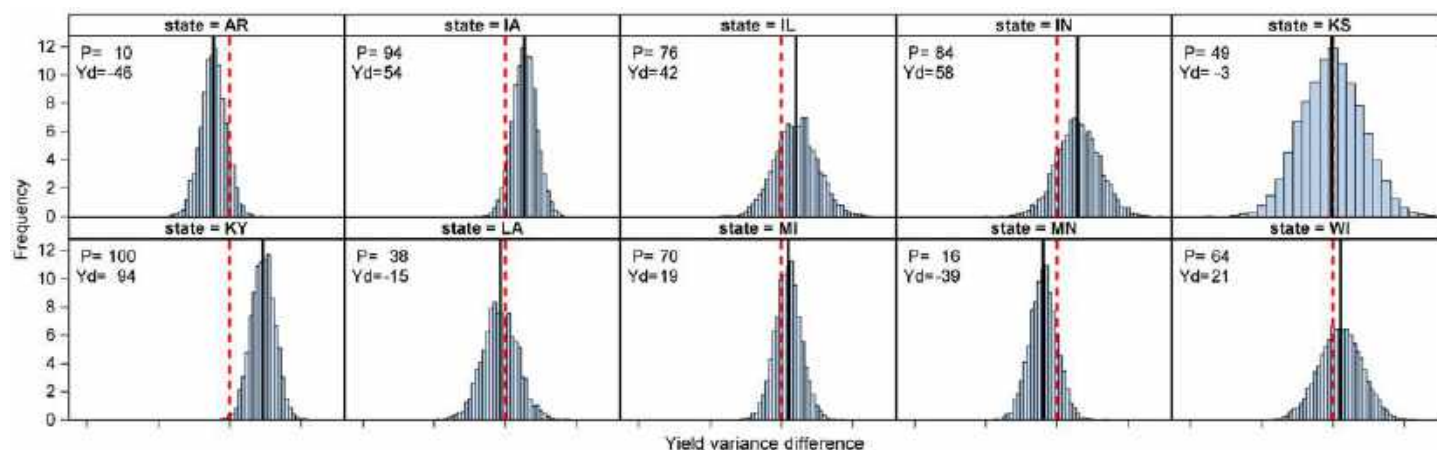


Figure 4. Distribution of yield skewness and the probability as a percentage that yield skewness < 0 in the posterior sample distribution (n=5,100) in high (blue histogram P(high)) and low (red histogram P(low)) input cropping systems in each state. Within each state, the red dashed line shows the zero-yield skewness.

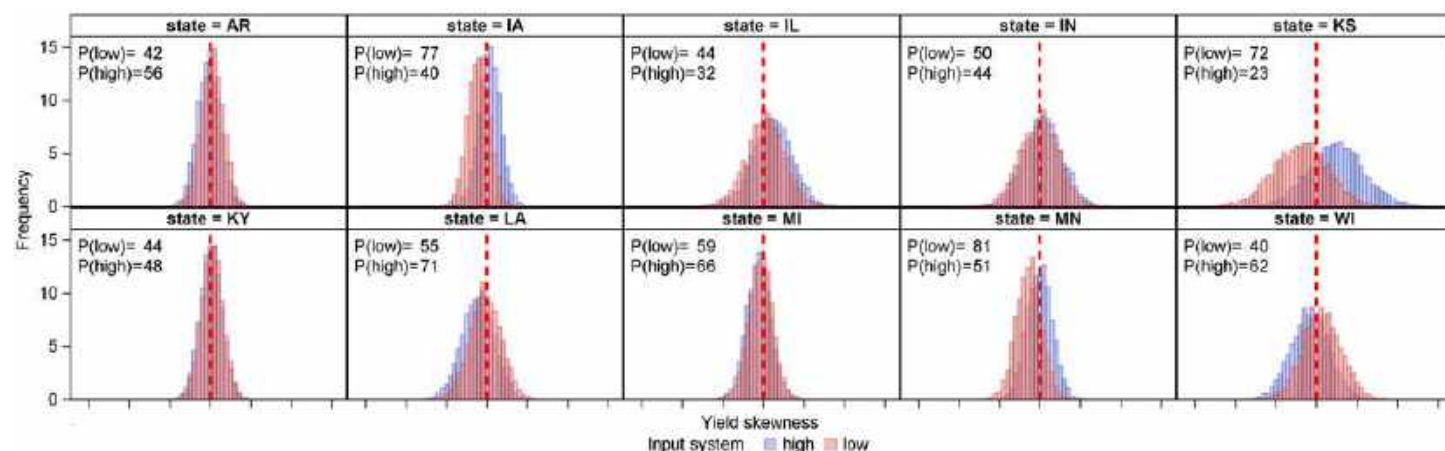


Figure 5. Distribution of yield kurtosis and the probability as a percentage that yield kurtosis > 0 in the posterior sample distribution (n=5,100) in high (blue histogram P(high)) and low (red histogram P(low)) input cropping systems in each state. Within each state, the red dashed line shows the zero-yield kurtosis.

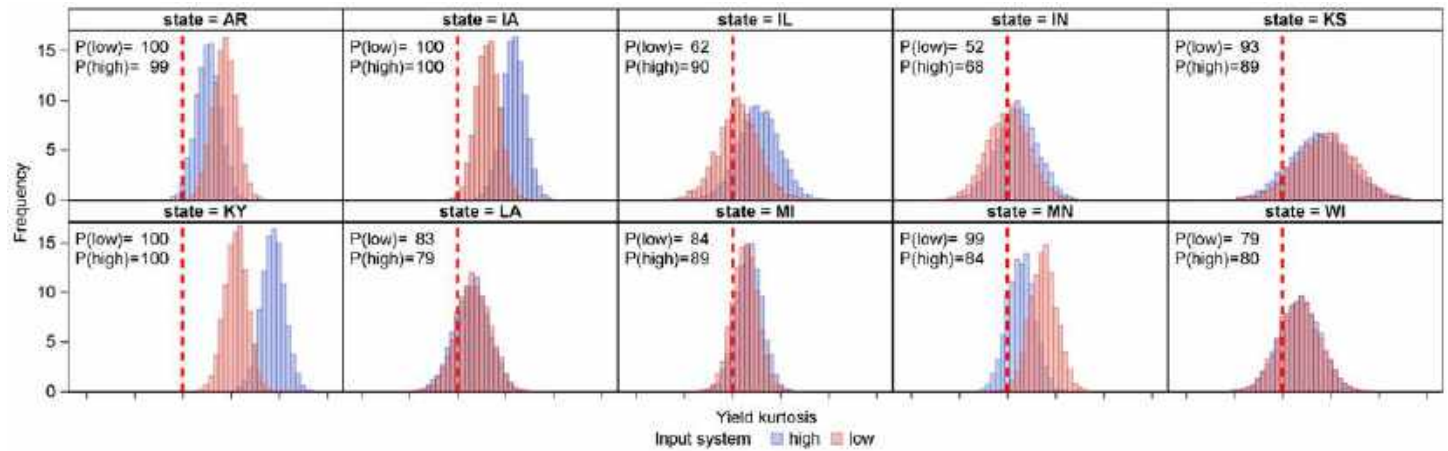
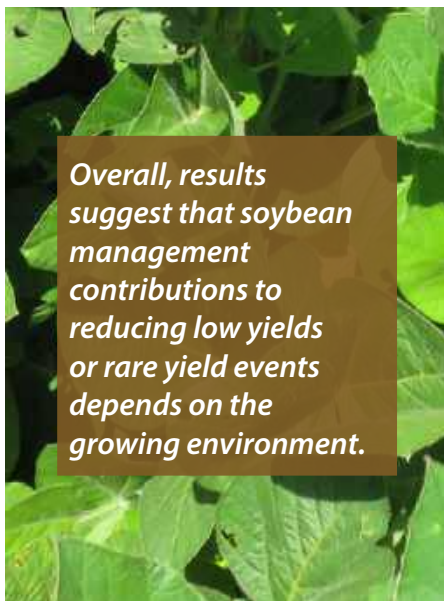
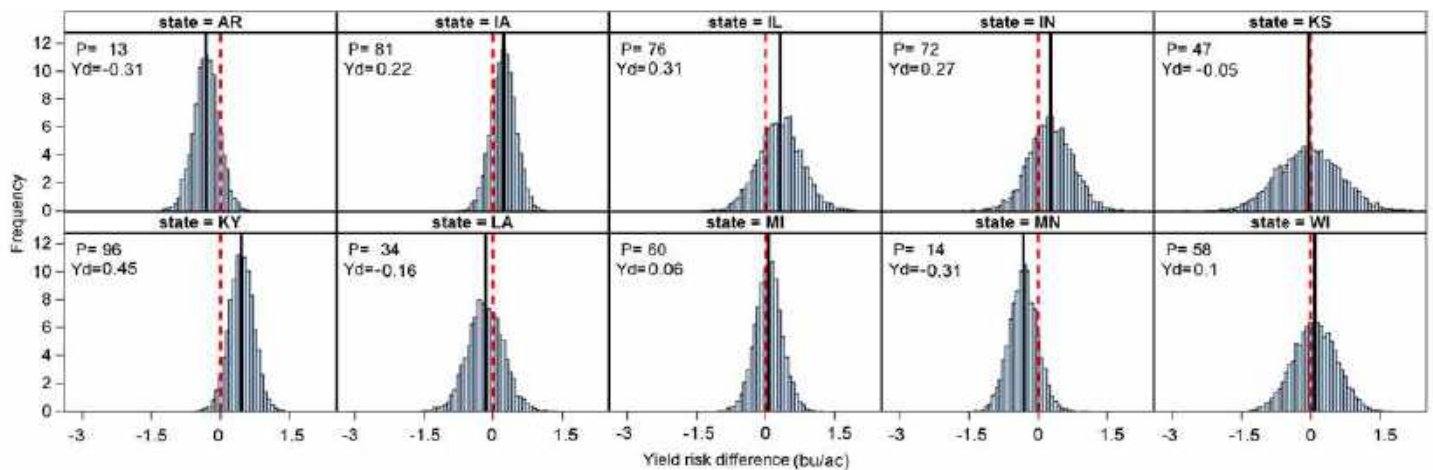


Figure 6. Distribution of the change in the cost of yield risk (bu/ac) between high- and low-input cropping systems in each state and probability (P) as a percentage that the cost difference > 0 in the posterior sample distribution (n=5,100). Within each state, the red dashed line shows the zero cost difference, and the black line indicates the mean cost difference (Yd).



Overall, results suggest that soybean management contributions to reducing low yields or rare yield events depends on the growing environment.

states, apart from KS where the two systems exhibited similar yield variance. Apart from KY, where the variance for low-input system was decreased, the observed differences were not significant. These results show that intensive soybean management does not consistently reduce yield risk across the examined growing environments.

Next, we examine differences in the yield skewness and kurtosis between systems. For yield skewness, high-input systems slightly increased yield skewness in IA, IL, IN and KS (Figure 4), implying lower risk of unusually low yields, while in AR, LA and WI, the yield skewness decreased. However, the effects were not significant, suggesting little empirical support that yield skewness changes significantly when using high-input management systems. For yield kurtosis, high-input systems exhibited lower kurtosis than low-input systems in AR, KS, LA and MN, suggesting lower risk exposure to the tails of the distribution (Figure 5). The opposite was observed in IA, IL, IN, KY and MI and no clear evidence of kurtosis differences were observed in the remaining states. Again, apart from AR, IA, KY and low-input system in MN, the results provide little empirical support that yield kurtosis changes significantly when using high-input management systems. Overall, results suggest that soybean management contributions to reducing low yields or rare yield events depends on the growing environment.



To evaluate the importance of the observed effects of management intensity on the yield distribution, we calculated the cost of risk (measured as bu/ac of soybean yield) by assuming a moderately risk averse and a highly risk averse farmer. For a moderately risk averse farmer ($\alpha=3$), the change in the total cost of risk between the high- and low-input systems was minimal, varying between -0.31 to 0.45 bu/ac (Figure 6) with no evidence that these differences were significant. The estimated mean costs of risk were less than 3% of their respective system-state-average yield.

DISCUSSION

U.S. farmers have become more interested in using multiple inputs in their soybean systems due to increased soybean prices, coupled with hypothesized nutrient deficiencies, potential plant health benefits, perceived level of potential pest infestation, and industry promotion. As an insurance practice, farmers typically decide to apply many of these inputs well in advance of planting time when there is little to no information about the potential severity or even presence of yield limiting factors. Here we detected a yield increasing effect due to high-input systems which was inconsistent among the examined states (-2.8 to 8.6 bu/ac or -4.9 to 12.7% of average yield).

Reducing the risk of crop failure and yield loss is desirable to farmers who rely on high and stable crop yield to ensure consistent profitability and business welfare. Our risk analysis showed that, the mean cost of the yield risk difference at the field level between high- and low-input systems in each state was minimal ($< 3\%$ of total yield) and inconsistent, suggesting that applying multiple inputs had little effect on reducing downside yield risk at the field level. Consistent with previous studies (Shi et al., 2013), we found that most of the cost of risk at the field level comes from the variance component and intensive management resulted in similar or increased variance compared to low-input systems. This result is consistent with the risk increasing effect of pest management inputs noted in previous studies (Horowitz and Lichtenberg, 1994; Salazar and Rand, 2020). The effect of management intensity on field-level yield skewness and kurtosis showed non-significant and inconsistent effects across the states. That the high-input system used multiple inputs, some protective and some productive, may have contributed to finding inconsistent effects. Our analysis shows that intensive management increased mean soybean yield and may increase the variance of yield at the field-level for many farmers, with no consistent effect on downside risk as measured by the higher moments of yield.

This analysis was at the field level for soybean production in isolation, but at the whole farm level, aggregate risk effects of high and low-input soybean systems can be different. Factors that may increase risk at the field level (such as high-input soybeans), can reduce risk in aggregate for a farm (Tack and Yu, 2020). For example, Hurley et al. (2004) showed that adopting Bt corn increased risk at the field level by increasing yield variance, but at the whole farm level, can increase or decrease farm risk depending on technology cost and how the farm adjusts the total area cropped. Also, farmers choose inputs to manage more than just yield or income risk, such as human and environmental safety or time and managerial simplicity (Hurley and Mitchell, 2020). Overall, these results suggest that the yield risk effects of intensive soybean management systems at the field level are not large and such systems should not necessarily be considered a broad-scale profitable and sustainable food-securing practice across the U.S. to reduce downside production loss risk. This is in agreement with recent work that showed that most of the management practices in the high-input system exhibited low-to-moderate importance in predicting soybean yield in major crop producing regions in the U.S. (Shah et al., 2021).

The greatest yield benefits from high-input systems were observed in the northern states of MN and WI (8 and 8.6 bu/ac, respectively). These mean yield increases accounted for a respective 12.7 and 12.1% of average yield and can be considered as substantial. The multiple applied inputs are estimated to cost ca. 170 \$/ac which means that the soybean price would need to exceed 20 \$/bu to cover the cost of application in MN and WI, respectively. Such high prices have never been observed



and therefore, these results question the profitability of broad scale adoption of high-input soybean management across the US. We argue that identifying management with high potential to consistently increase yield in specific regions will be more likely to result in profitable yield increases (Andrade et al., 2022).

CONCLUSIONS

Overall results in this work show that when compared to low input application, intensifying soybean management is yield increasing without reducing downside yield loss risk. Additionally, the observed yield benefits indicate a negative return on investment which is consistent with previous studies (Orlowski et al., 2016; Quinn & Steinke, 2019). These results further support the use of integrated pest management (IPM) for making input decisions instead of relying on prophylactic input applications as insurance against yield-limiting factors. Such approach can be cost-effective and environmentally friendly since inputs are applied when and where needed. We conclude that future studies of food security and crop production should be region-specific and focus on identifying management practices with the greatest yield potential based on IPM practices rather than recommending broad-scale intensive management systems as insurance practice.

Adapted from: Mourtzinis, S., Mitchell, P., Esker, P., Cerrudo, A., Naeve, S., & Conley, S. 2023. *Field-level yield benefits and risk effects of intensive soybean management across the U.S.*

REFERENCES

- Andrade, J. F., Mourtzinis, S., Rattalino Edreira, J. I., Conley, S. P., Gaska, J. M., Kandel, H. J., Lindsay, L. E., Naeve, S., Nelson, S., Sigh, M., Thompson, L., Specht, J. E. & Grassini, P. *Field validation of a farmer-data approach to close soybean yield gaps in the US North Central region.* *Agric. Syst.* 200, 103434 (2022).
- Alexandratos, N. & Bruinsma, J. *World Agriculture Towards 2030/2050: the 2012 Revision.* FAO, Rome (2012).
- Orlowski, J.M., Haverkamp, B.J., Laurenz, R.G., Marburger, D.A., Wilson, E.W., Casteel, S.N., Conley, S.P., Naeve, S.L., Nafzinger, E.D., Roozeboom, K.L., Ross, W.J., Thelen, K.D., & Lee, C.D. *High-input management systems effect on soybean seed yield, yield components, and economic break-even probabilities.* *Crop Sci.* 56:1988–2004 (2016). doi:10.2135/cropsci2015.10.0620
- Quinn, D. & Steinke, K. *Comparing high- and low-input management on soybean yield and profitability in Michigan.* *Crop Forage Turfgrass Manage.* 5:190029 (2019). doi:10.2134/cftm2019.04.0029
- Tilman, D., Balzer, C., Hill, J., & Befort, B.L. *Global food demand and the sustainable intensification of agriculture.* *Proc. Natl. Acad. Sci. U. S. A.* 108, 20260–20264 (2011).
- Grassini, P., Eskridge, K., & Cassman, K.G. *Distinguishing between yield advances and yield plateaus in historical crop production trends.* *Nat. Commun.* 4, 2918 (2013).
- Marburger, D.A., Haverkamp, B.J., Laurenz, R.G., Orlowski, J.M., Wilson, E.W., Casteel, S.N., Lee, C.D., Naeve, S.L., Nafzinger, E.D., Roozeboom, K.L., Ross, W.J., Thelen, K.D., & Conley, S.P. *Characterizing genotype $\times$ management interactions on soybean seed yield.* *Crop Sci.* 56:786–796 (2016). doi:10.2135/cropsci2015.09.0576
- Shi, G., Chavas, J.P., & Lauer, J. *Commercialized transgenic traits, maize productivity and yield risk.* *Nat. Biotechnol.* 31, 111–114 (2013). <https://doi.org/10.1038/nbt.2496>
- Horowitz, J.K. & Lichtenberg, E. *Risk-reducing and risk-increasing effects of pesticides.* *J. Agric. Econ.* 45(1) 82–89 (1994).
- Salazar, C. & Rand, J. *Pesticide use, production risk and shocks. The case of rice producers in Vietnam.* *J. Environ. Manage.* 253 109705 (2020).
- Hurley, T.M., & Mitchell, P.D. *The value of insect management to US maize, soybean and cotton farmers.* *Pest Manage. Sci.* 76, 4159–4172 (2020). <https://doi.org/10.1002/ps.5974>
- Hurley, T.M., Mitchell, P.D., & Rice, M.E. *Risk and the value of Bt corn.* *Amer. J. Agric. Econ.* 86, 345–358 (2004).
- Tack, J., & Yu, J. *Risk management in agricultural production.* Barrett, C.B., & Just, D.R., eds., *Handbook of Agricultural Economics*, Vol 5, pp. 4135–4231 (2021). <https://doi.org/10.1016/bs.hesagr.2021.10.004>
- Shah, A. D., Butts, T. R., Mourtzinis, S., Rattalino Edreira, J. I., Grassini, P., Conley, S. P., & Esker, P. D. *An interpretable machine learning assessment of foliar fungicide contribution to soybean yield in the north-central United States.* *Scientific Reports* 11:18769 (2021).