

## 4. Krylov Methods

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## Reference

J.R. Shewchuk:

### An Introduction to the Conjugate Gradient Method Without the Agonizing Pain

Describes **Steepest Descent** and **CG** (at least with not so much pain).



# Steepest Descent - Algorithm



## Steepest Descent - Algorithm

- Choose

$$\vec{r} = \vec{b} - A\vec{u}$$

as search direction if the current approximation is  $\vec{u}$ ,

- Choose

$$\alpha = \frac{\vec{r}^T \vec{r}}{(A\vec{r})^T \vec{r}}$$

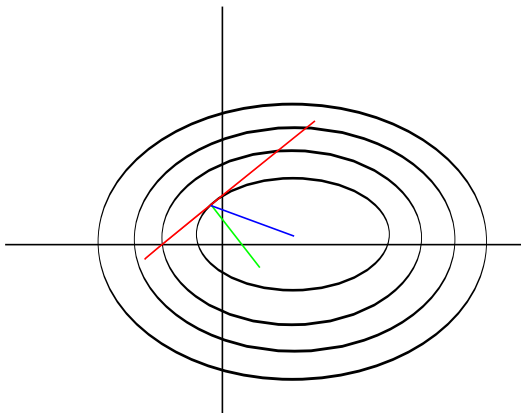
as stepsize,

- New approximation:

$$\vec{u}^{new} = \vec{u} + \alpha \vec{r}.$$



## Conjugate Gradients - Idea



We could make the individual search directions be orthogonal to each other. Then, we end up with a direct solver. However, we should not use the standard orthogonality (green), but vector product induced by  $A$  (blue).

## Conjugate Gradients - Algorithm

How far to go along  $d_i$ ?  $\alpha = \frac{r_i^T d_i}{d_i^T A d_i}$

Compute new  $u_i$   $u_{i+1} = u_i + \alpha d_i$

Compute new residual  $r_{i+1} = r_i - \alpha A d_i$

Compute  $\beta$ , so that  $\beta = \frac{r_{i+1}^T r_{i+1}}{r_i^T r_i}$

$d_i$  and  $d_{i+1}$  are orthogonal  $d_{i+1} = r_{i+1} - \beta d_i$